

Self-Driving Networks: Use Cases, Approaches, and Research Challenges

Stefan Schmid

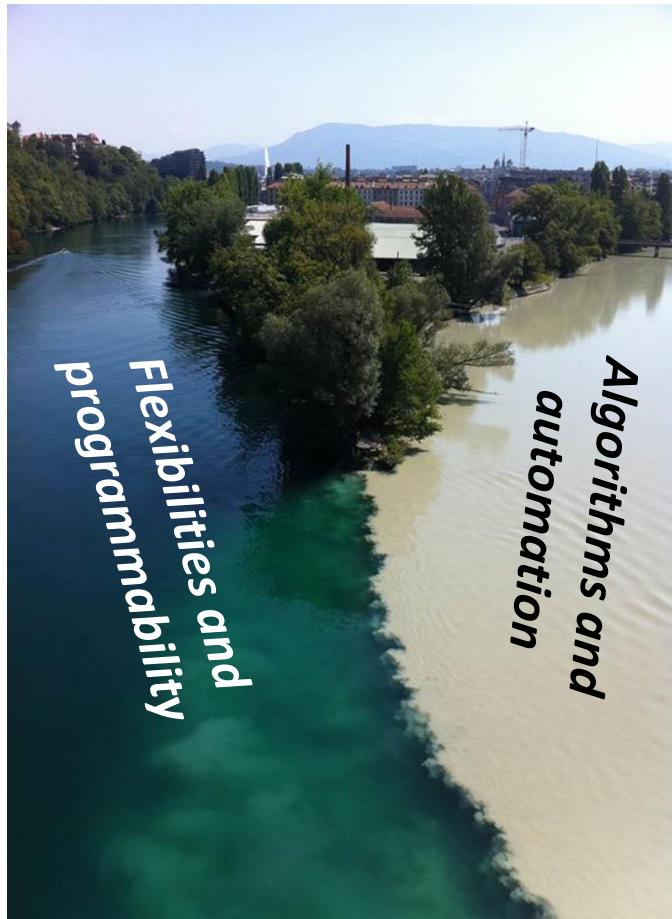
“We cannot direct the wind,
but we can adjust the sails.”

(Folklore)

Acknowledgements:



It's a Great Time to Be a Networking Researcher!

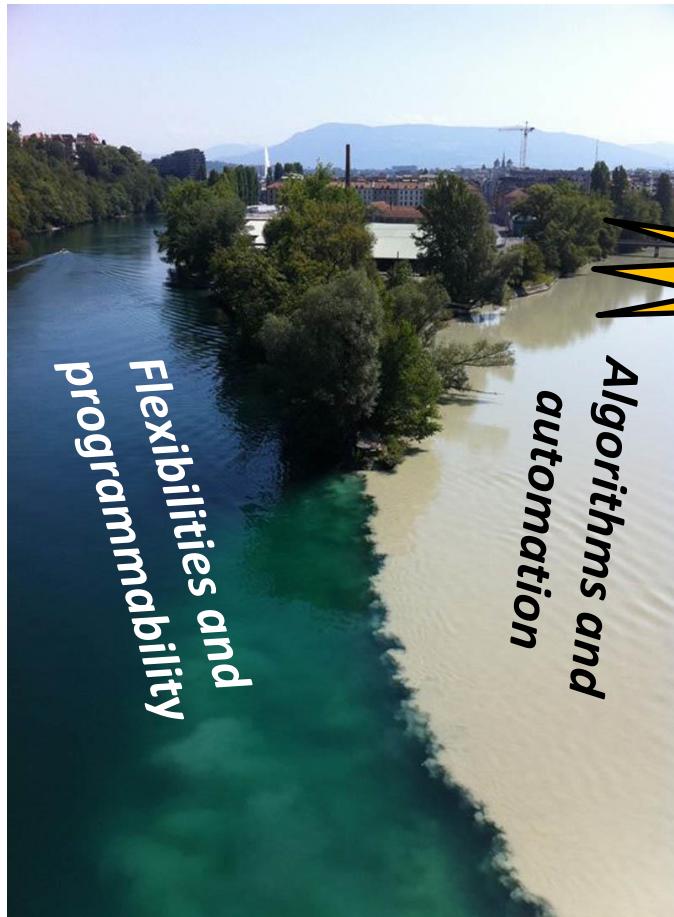


*Algorithms and
automation*

*Flexibilities and
programmability*

Rhone and Avre (Switzerland)

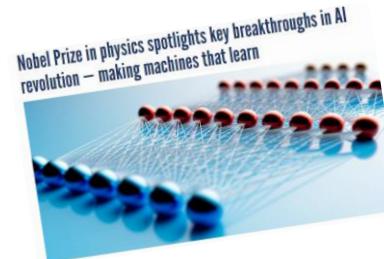
It's a Great Time to Be a Networking Researcher!



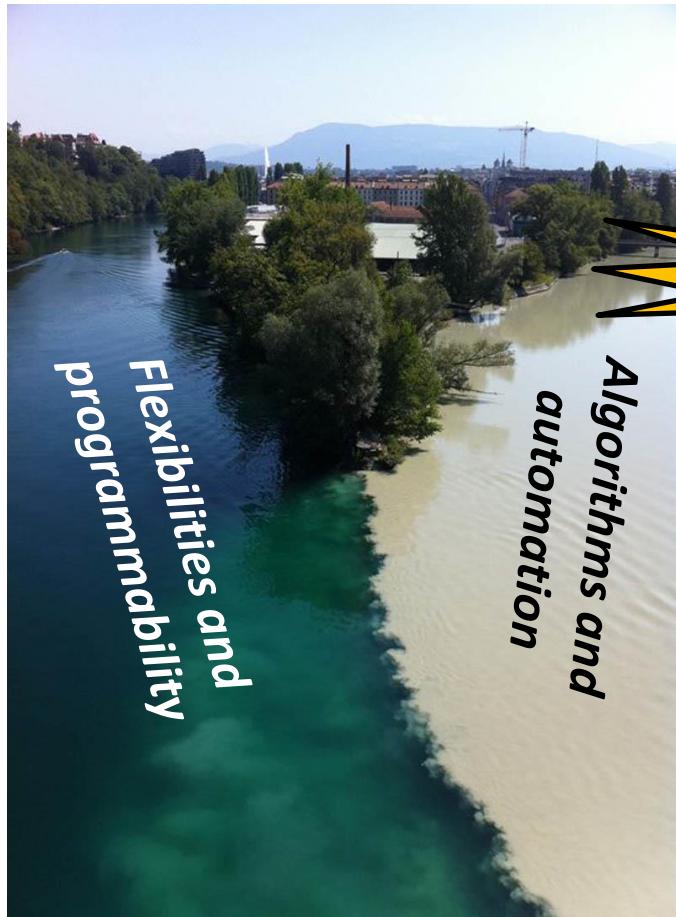
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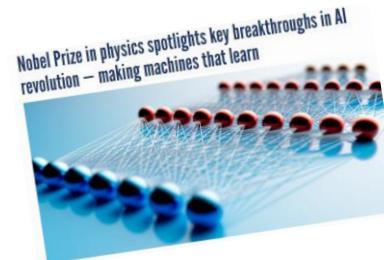
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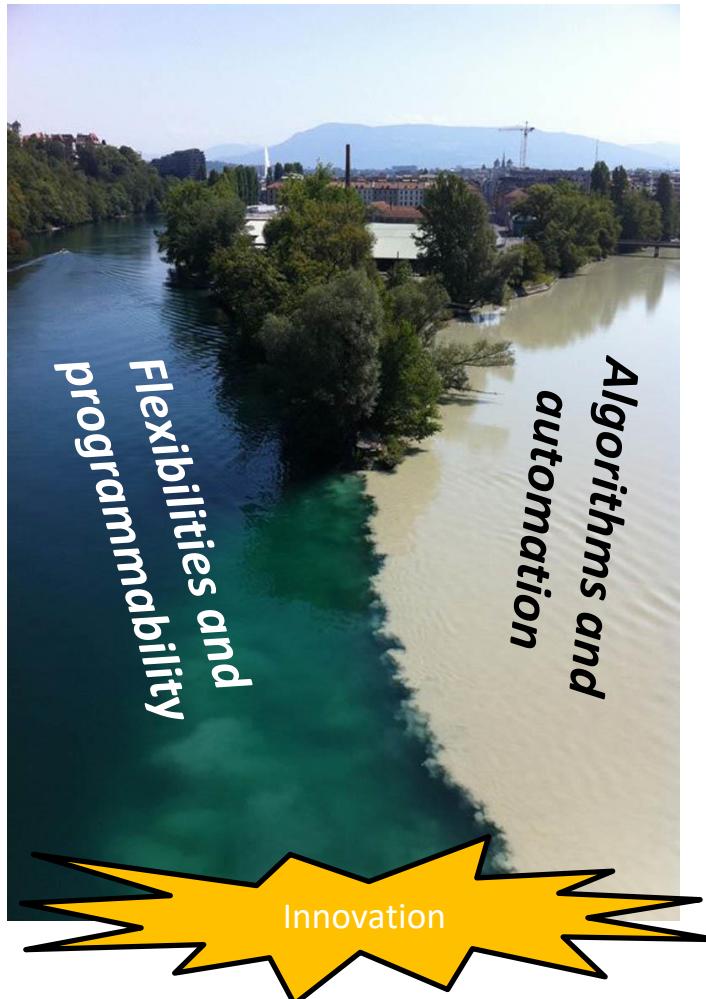
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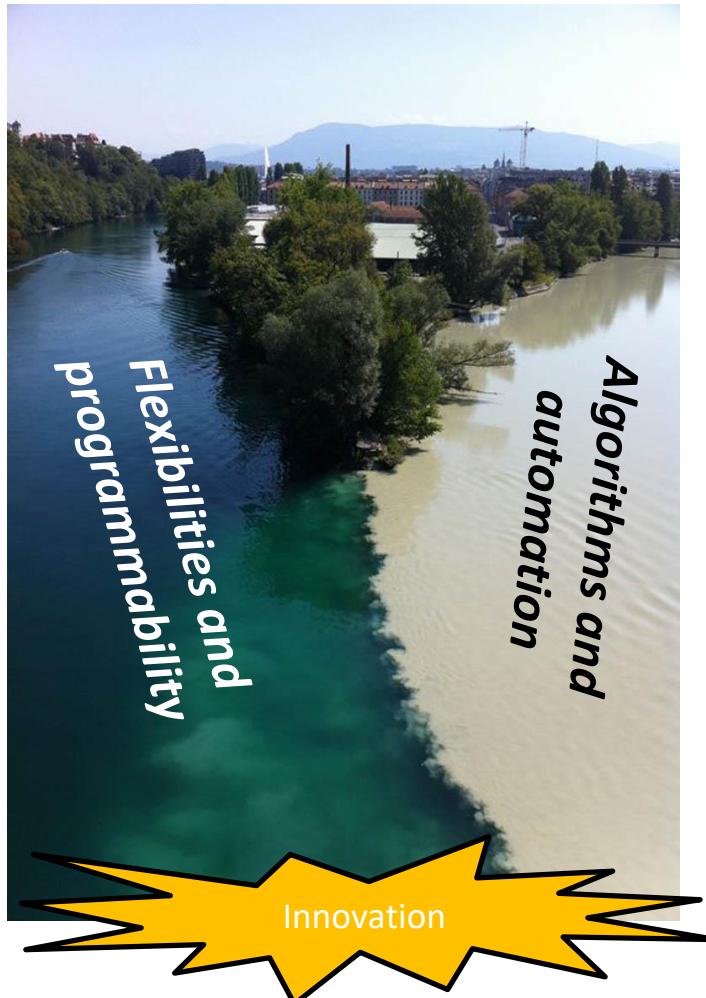
Economics and literature?



It's a Great Time to Be a Networking Researcher!



It's a Great Time to Be a Networking Researcher!



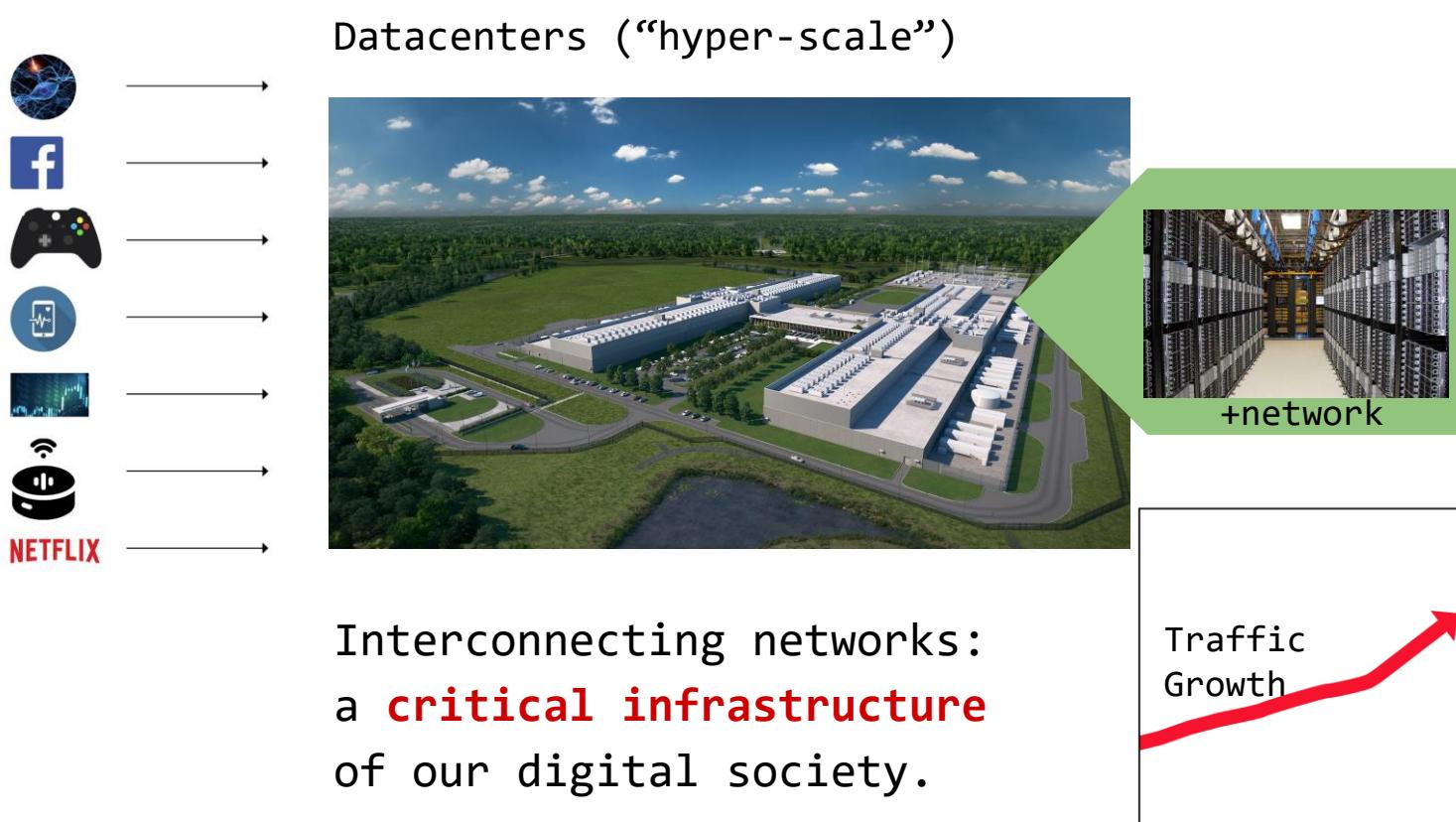
Credits: George Varghese

Enables and motivates
self-driving networks!



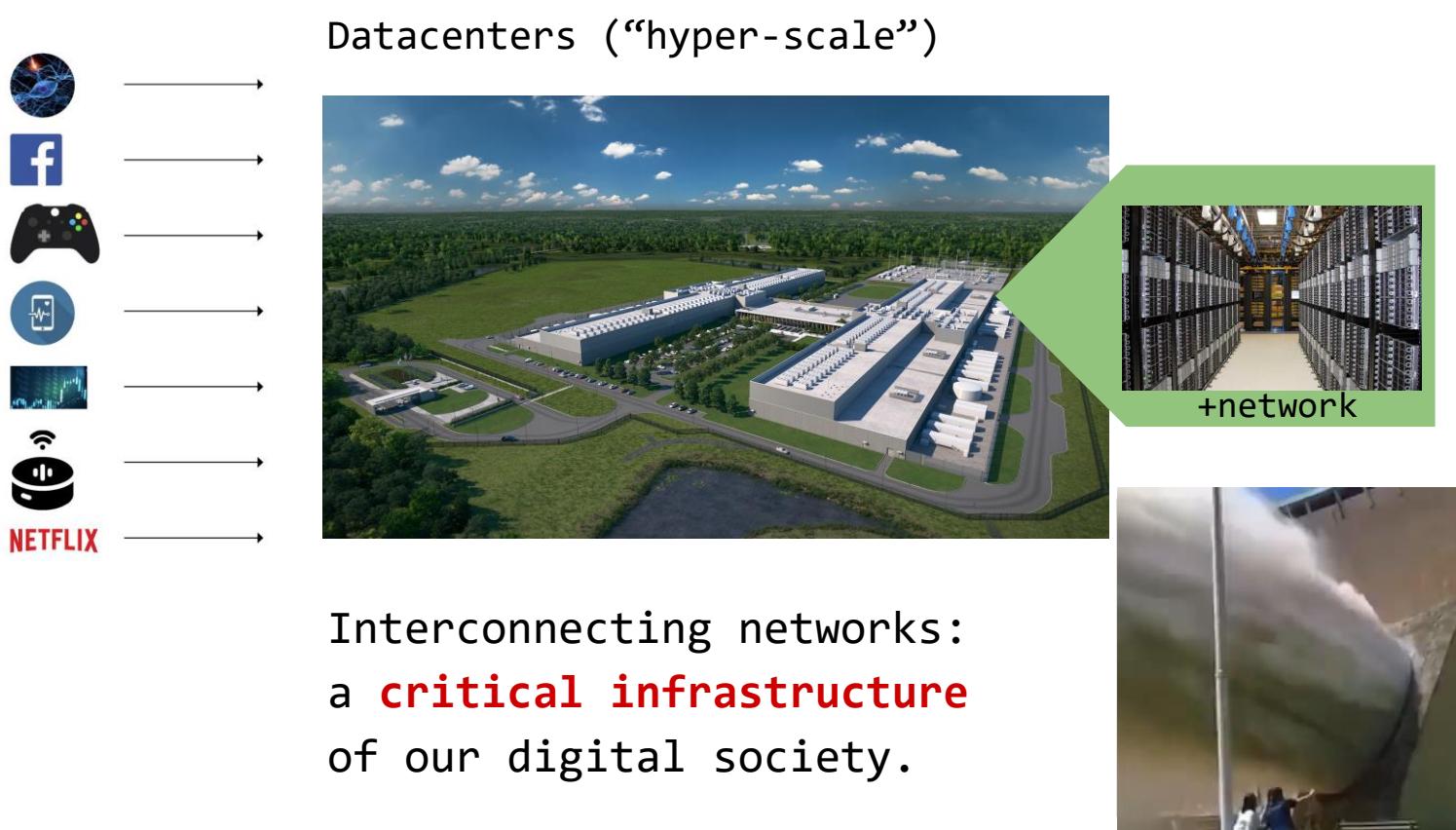
Innovations Needed!

Explosive Traffic



Innovations Needed!

Explosive Traffic



Credits: Marco Chiesa

Fast growing traffic also in...

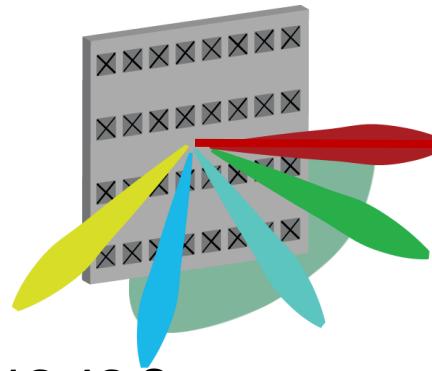
... wireless and mobile



From generation to generation more...

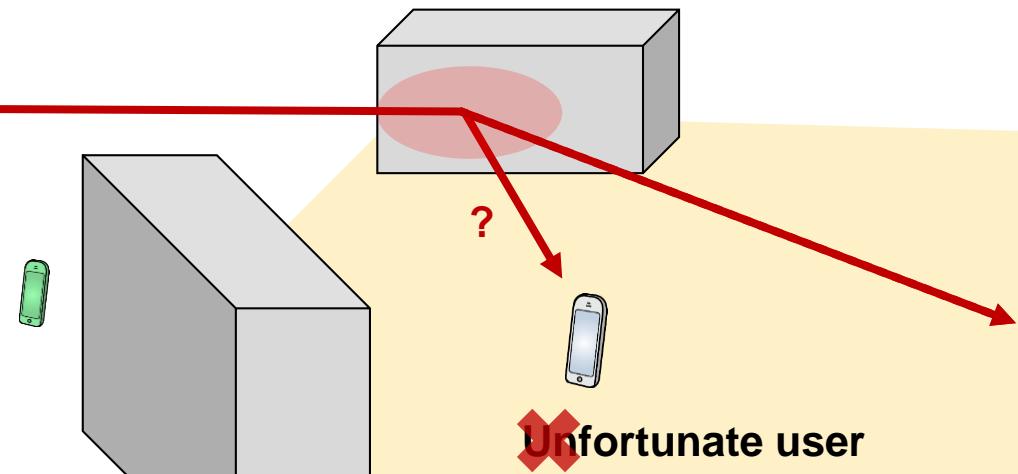
Exciting Flexibilities

5G: Adaptive multi-user beamforming



1G-4G Sector antenna
Fixed radiation pattern

6G: Control objects in the environment?

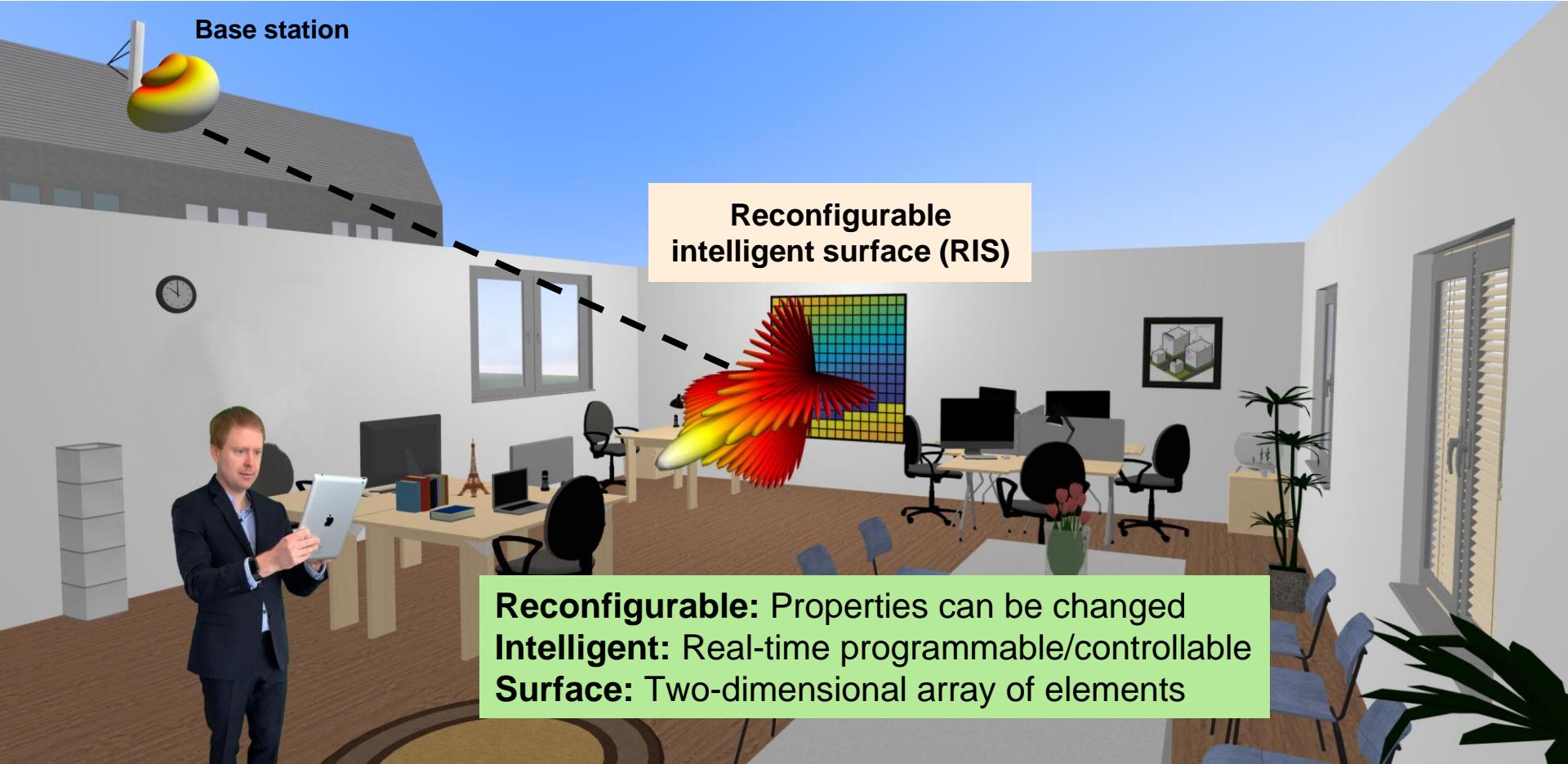


Fortunate user

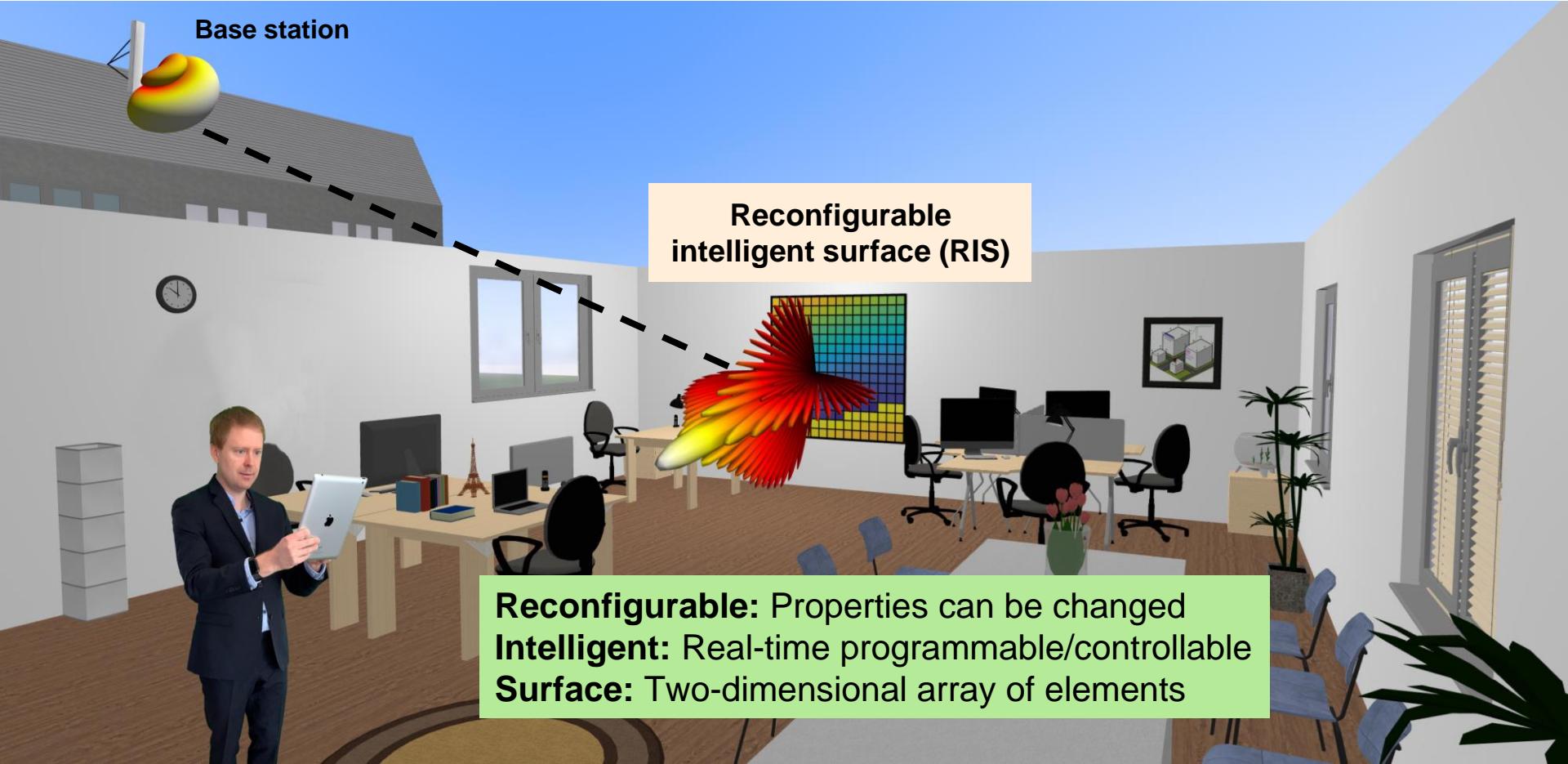
Traditionally limited by
Line of Sight Only



Reconfigurable Intelligent Surfaces: Extend to Virtual Line of Sight



Reconfigurable Intelligent Surfaces: Extend to Virtual Line of Sight



Great opportunities but come with...

Challenges

- With growing ***demand*** for networks, also increasing ***dependability***
- Important step toward dependable networks: ***modelling***...
- ... and ***automation*** (also using formal methods)!
- Contributions from the ICIN community critical

It's high time for computer-aided designs! Reality vs Requirements

Today, dependability requirements stand in contrast with reality:

Countries disconnected

Data Centre ▶ Networks

Google routing blunder sent Japan's Internet dark on Friday

Another big BGP blunder

By Richard Chirgwin 27 Aug 2017 at 22:35

40 SHARE ▾

Last Friday, someone in Google fat-thumbed a border gateway protocol (BGP) advertisement and sent Japanese Internet traffic into a black hole.

The trouble began when The Chocolate Factory "leaked" a big route table to Verizon, the result of which was traffic from Japanese giants like NTT and KDDI was sent to Google on the expectation it would be treated as transit.

Passengers stranded

British Airways' latest Total Inability To Support Upwardness of Planes* caused by Amadeus system outage

Stuck on the ground awaiting a load sheet? Here's why

By Gareth Corfield 19 Jul 2018 at 11:16

109 SHARE ▾



*A flight around the world was canceled as a result of the Amadeus outage

Even 911 affected

Officials: Human error to blame in Minn. 911 outage

According to a press release, CenturyLink told department of public safety that human error by an employee of a third party vendor was to blame for the outage

Aug 16, 2018

Duluth News Tribune

SAINT PAUL, Minn. — The Minnesota Department of Public Safety Emergency Communication Networks division was told by its 911 provider that an Aug. 1 outage was caused by human error.

Even tech-savvy companies struggle:



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Even tech-savvy companies struggle:



Mainly:
human
errors!

It's high time for computer-aided designs!

Reality vs Requirements

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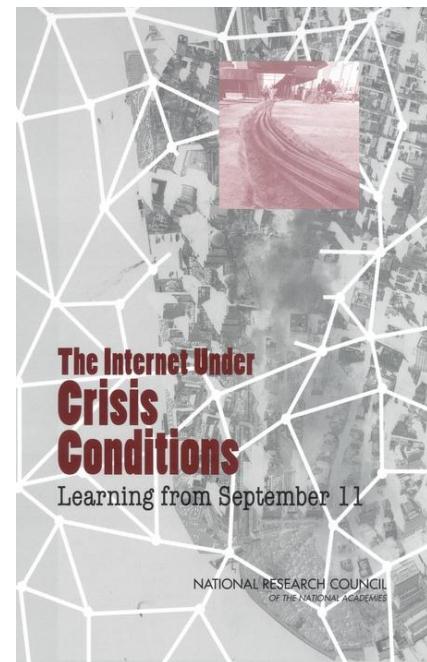
Even tech-savvy companies struggle:



Wireless particularly challenging to model!

An Anecdote

- Report by the National Research Council about ***9/11/2001 attacks***
- While the core Internet infrastructure installed in the WTC was down, the overall Internet was ***more stable*** than usual
- ... because operators stopped touching network devices?!



Roadmap

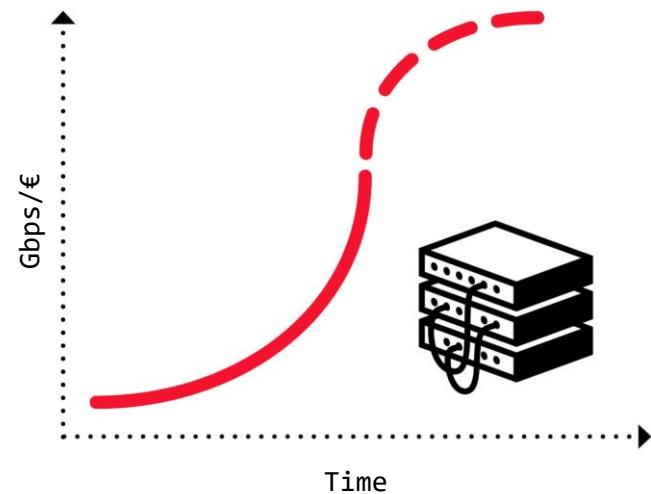


- Performance: Self-adjusting datacenter networks
- Modelling: How to model workloads, such as ML workloads?
- Dependability: Self-correcting MPLS networks
- More Use cases for self-driving networks

Datacenters Today

Huge Infrastructure, Inefficient Use

- Network equipment reaching capacity limits
 - Transistor density rates stalling
 - “End of **Moore’s Law** in networking”
- Hence: more equipment, larger networks
- Resource intensive and: **inefficient**



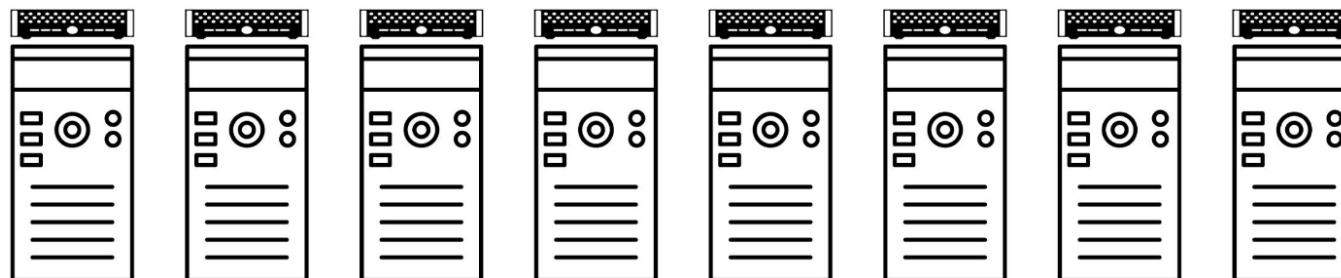
[1] Source: Microsoft, 2019

Annoying for companies,
opportunity for researchers!

Root Cause

Fixed and Demand-Oblivious Topology

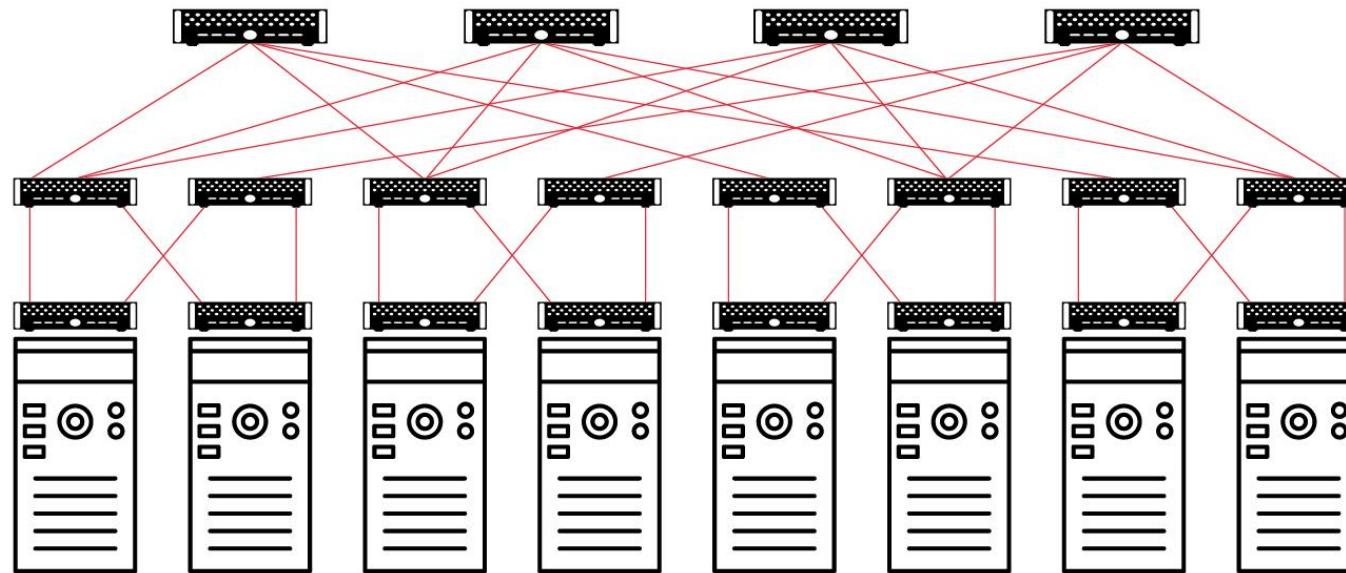
How to interconnect?



Root Cause

Fixed and Demand-Oblivious Topology

- Example: fat-tree topology (bi-regular)
 - 2 types of switches: top-of-rack (ToR) connect to hosts, additional switches connecting switches to increase throughput

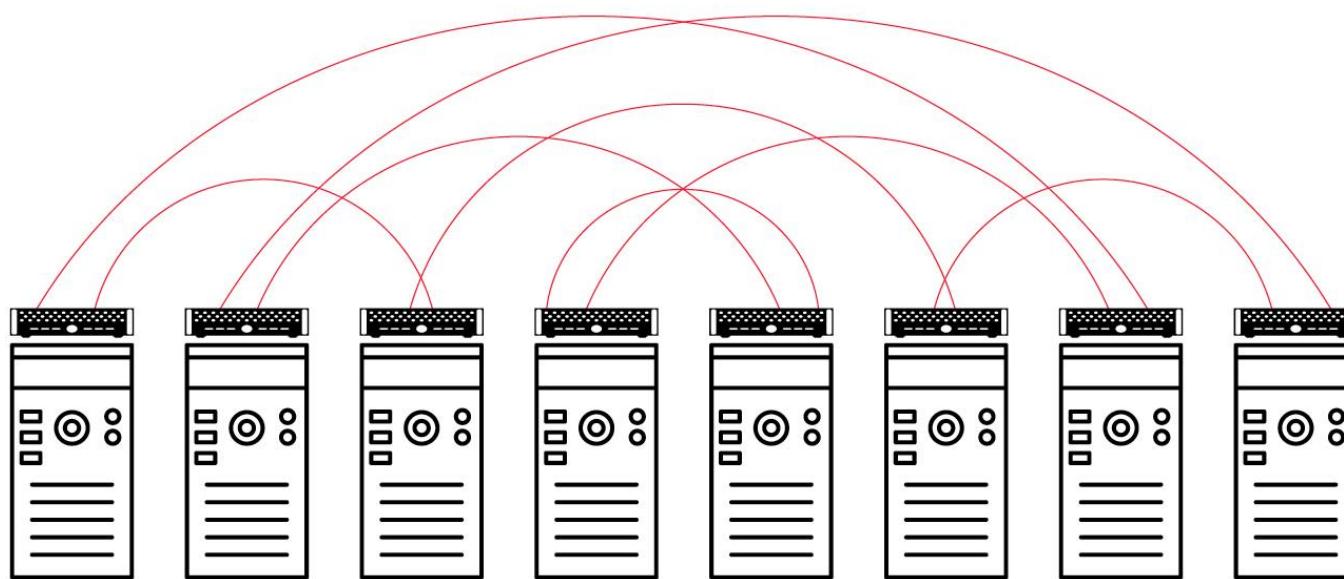


Root Cause

Fixed and Demand-Oblivious Topology

→ Example: expander topology (uni-regular)

→ Only 1 type of switches:
lower installation and management overheads



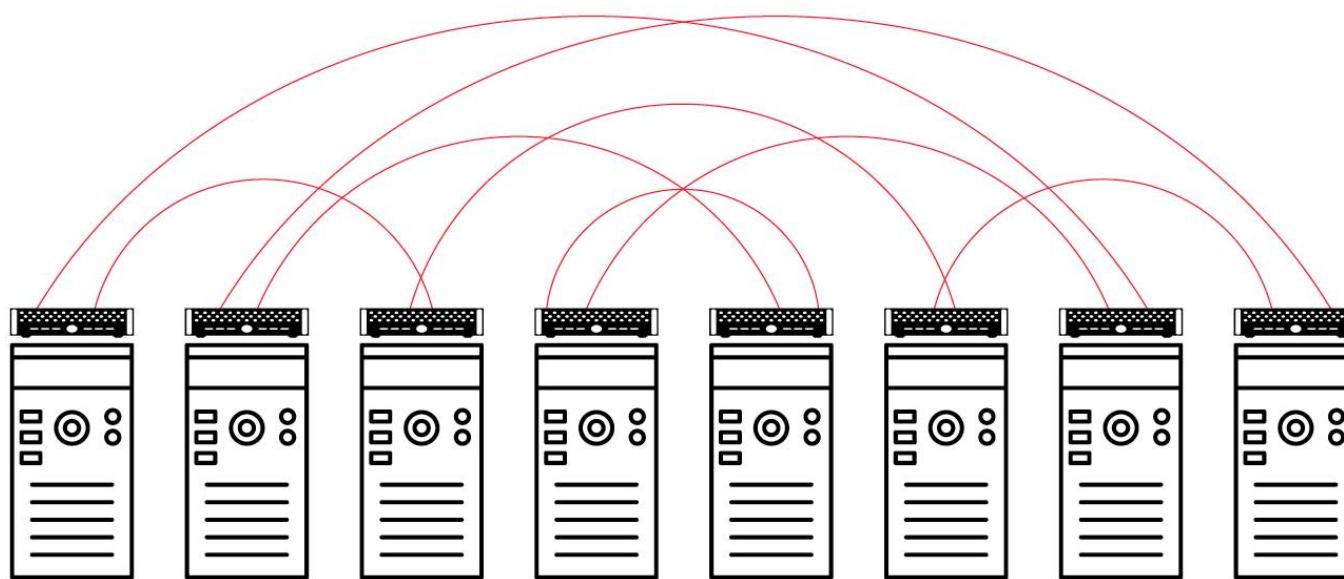
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Highway which ignores
actual traffic: **frustrating!**



Root Cause

Fixed and Demand-Oblivious Topology



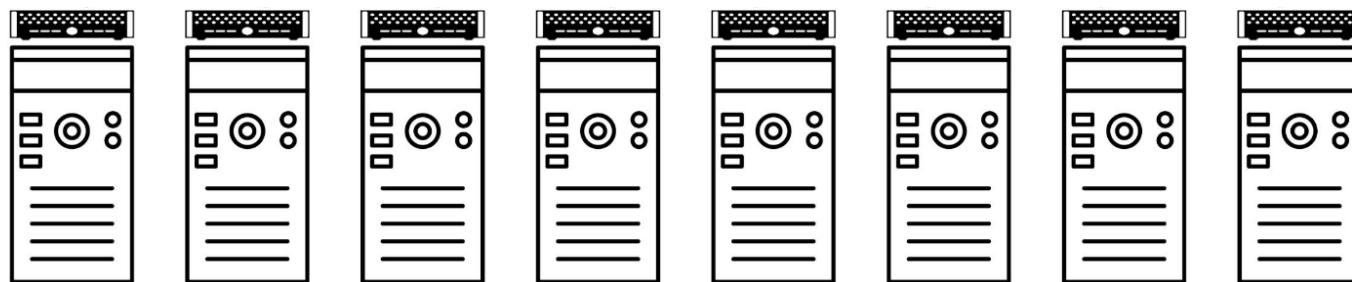
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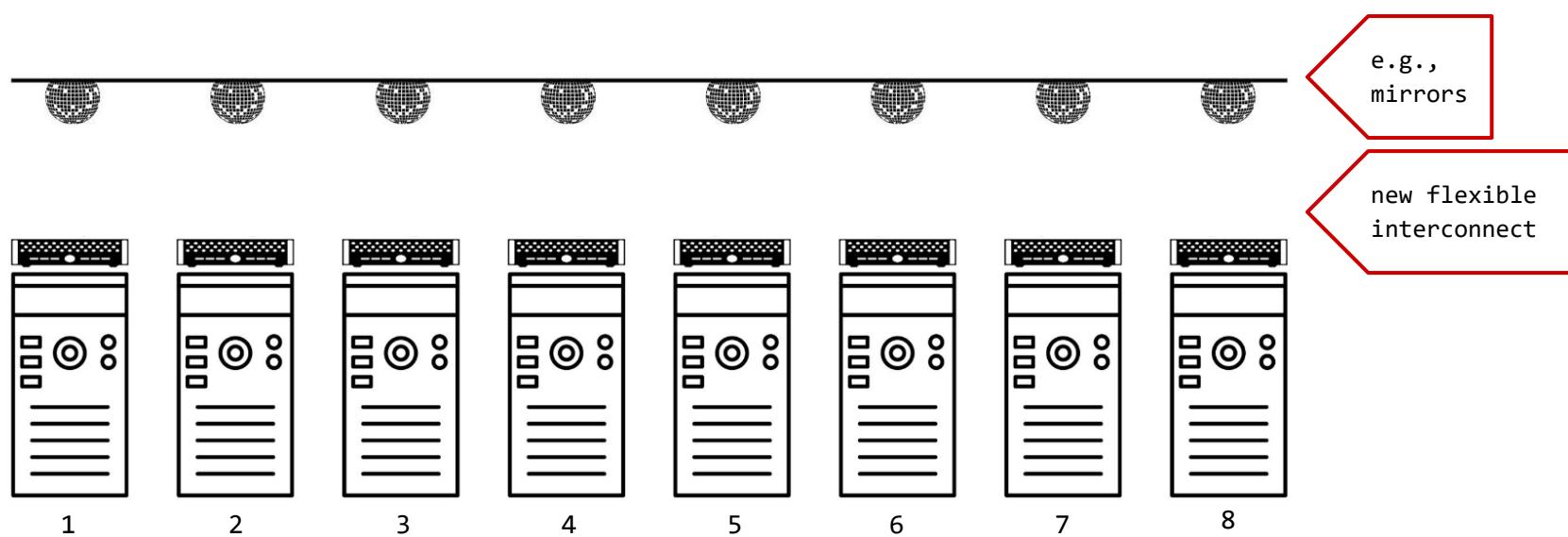
A Vision

Flexible and Demand-Aware Topologies



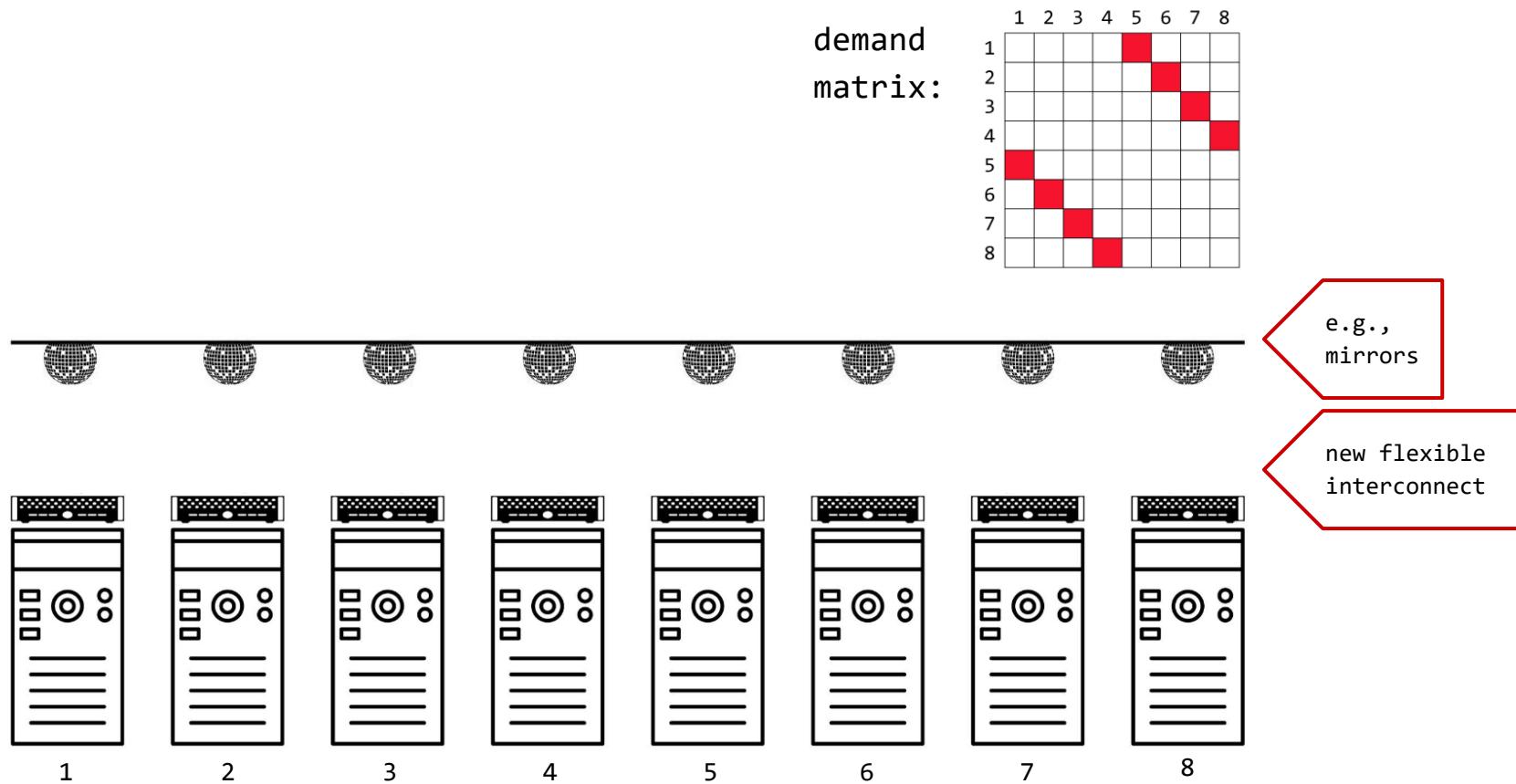
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Flexible and Demand-Aware Topologies



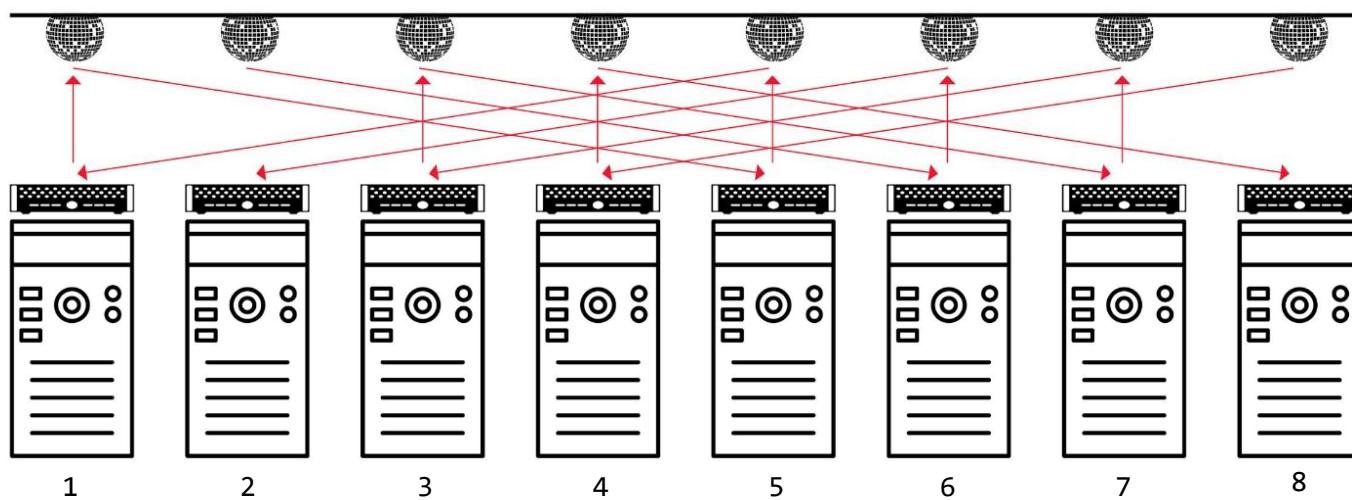
A Vision

Flexible and Demand-Aware Topologies

Matches demand

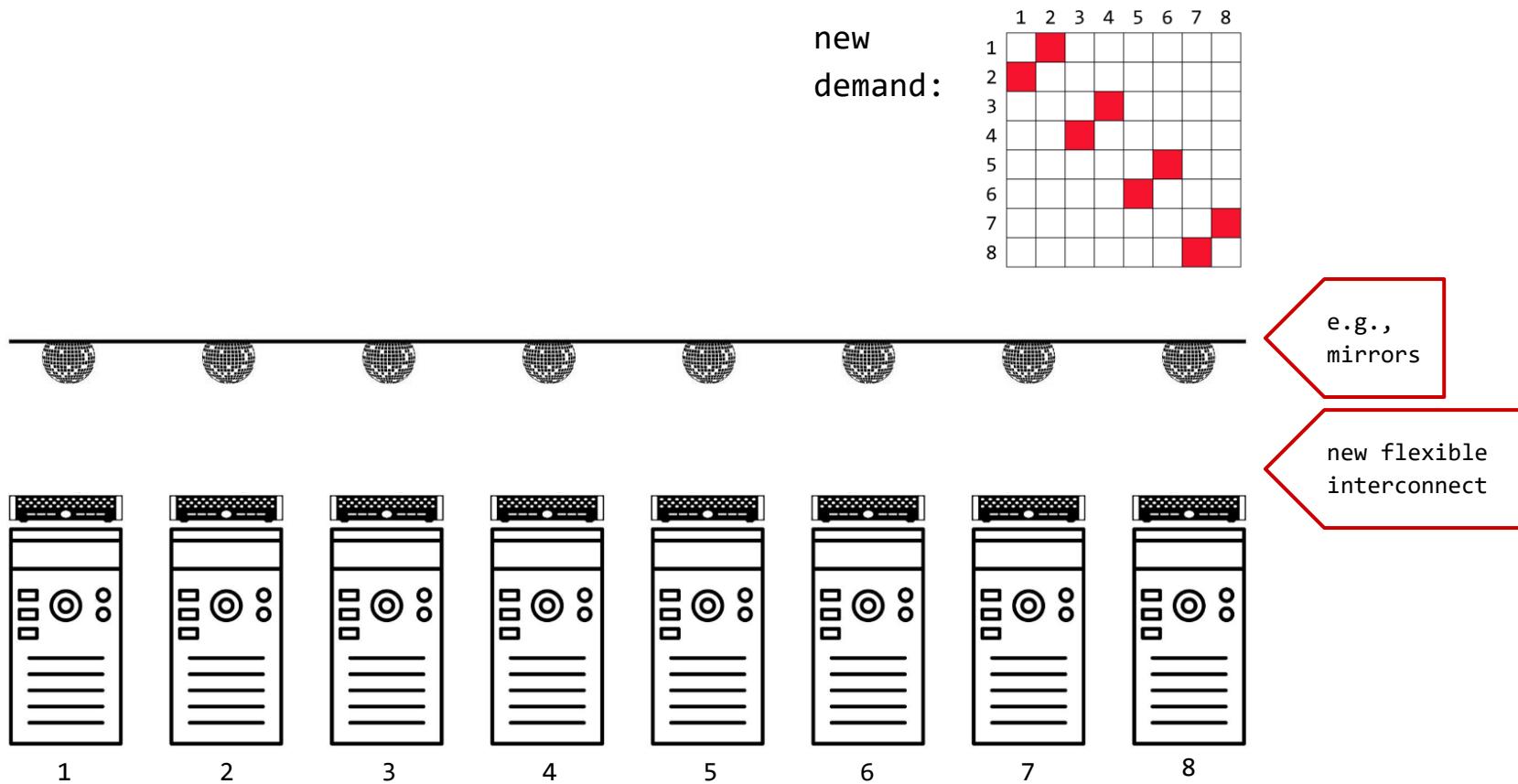
demand
matrix:

1	2	3	4	5	6	7	8
1					■		
2						■	
3							■
4							■
5	■						
6		■					
7			■				
8				■			



A Vision

Flexible and Demand-Aware Topologies



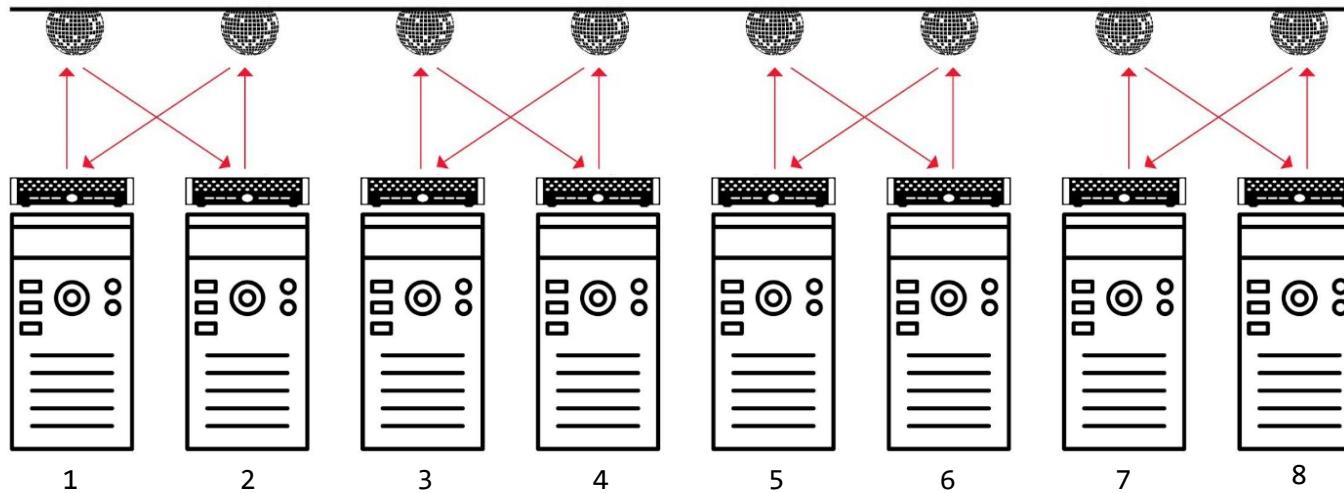
A Vision

Flexible and Demand-Aware Topologies

Matches demand

new
demand:

1	2	3	4	5	6	7	8
1							
2	■						
3							
4		■		■			
5							
6					■		
7						■	
8							■



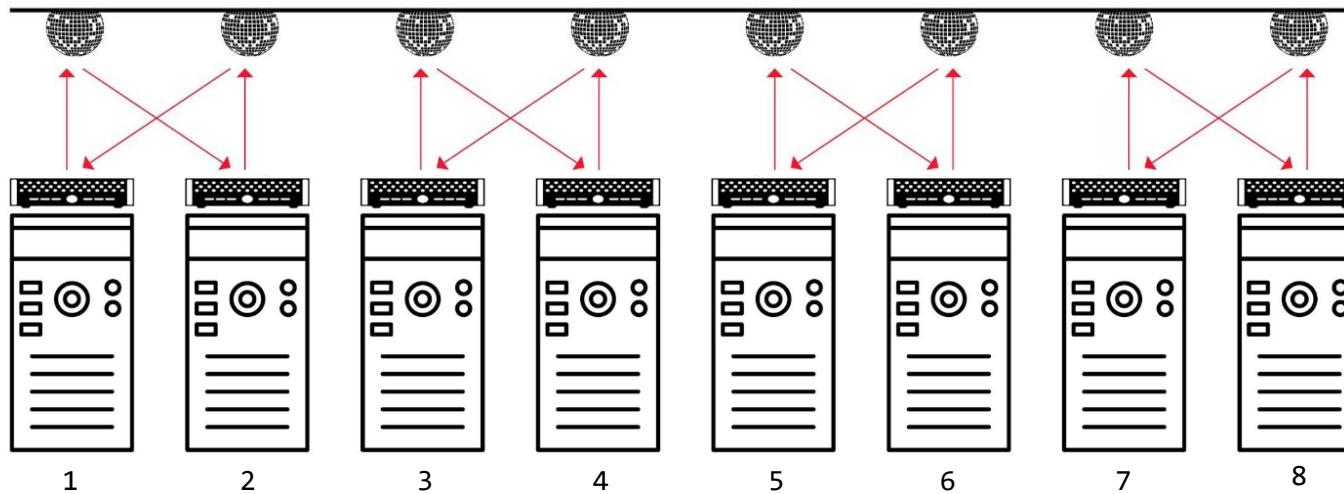
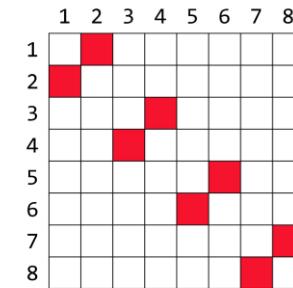
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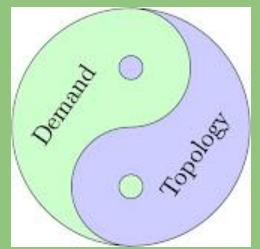


Self-Adjusting
Networks

new
demand:



“Yin and Yang”-Networking!

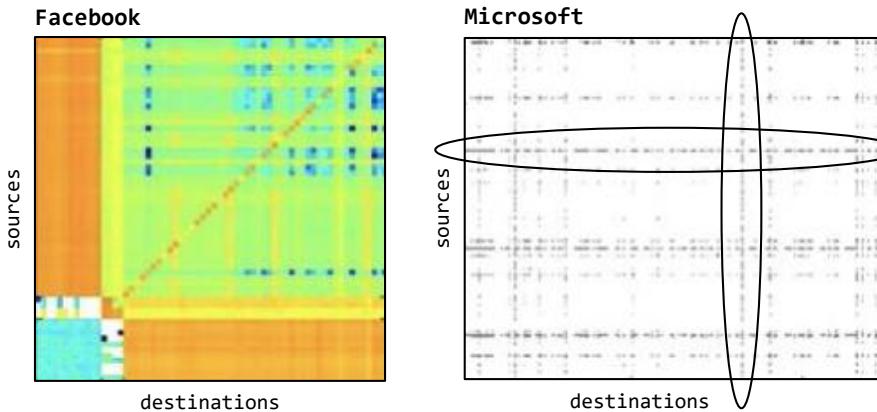


The Motivation

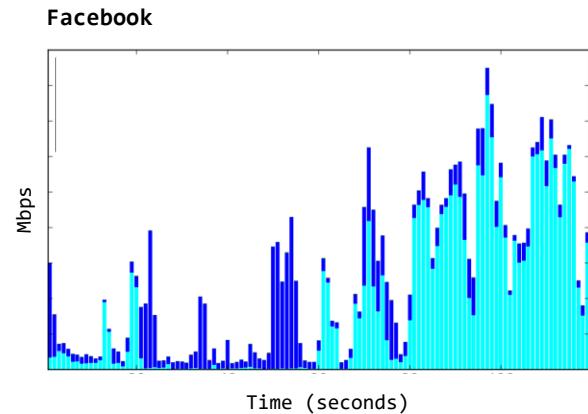
Much Structure in the Demand

Empirical studies:

traffic matrices **sparse** and **skewed**



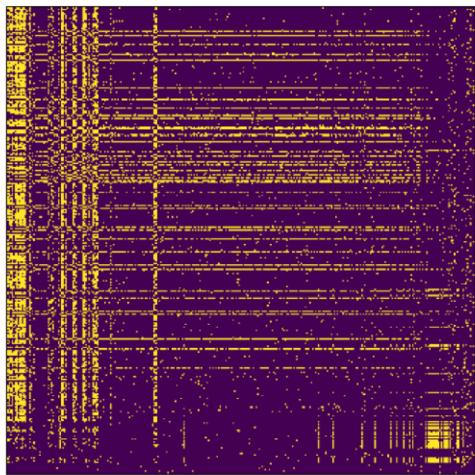
traffic **bursty** over time



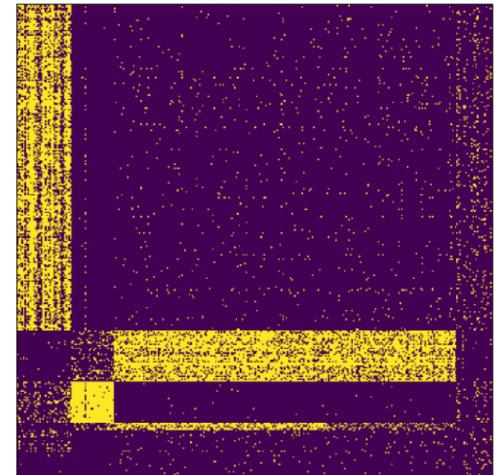
The **hypothesis**: can
be exploited.

Traffic is also clustered:

Small Stable Clusters

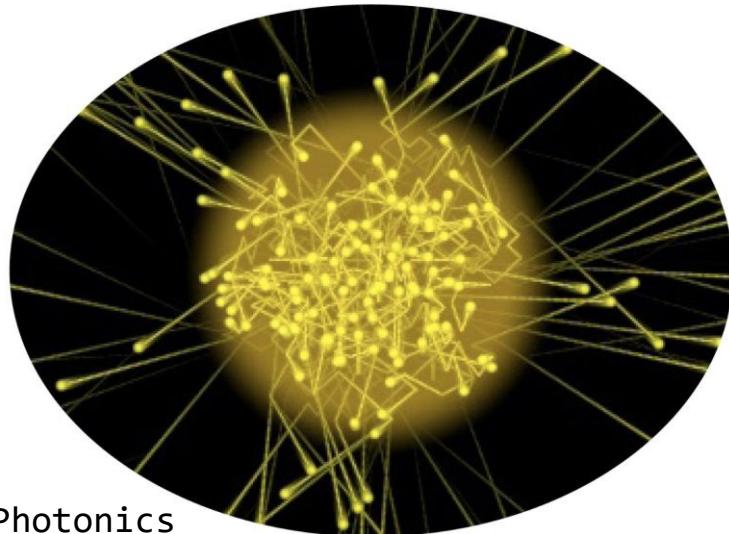


reordering based on
bicluster structure



Opportunity: *exploit* with little reconfigurations!

Sounds Crazy? Emerging Enabling Technology.



H2020:

**“Photronics one of only five
key enabling technologies
for future prosperity.”**

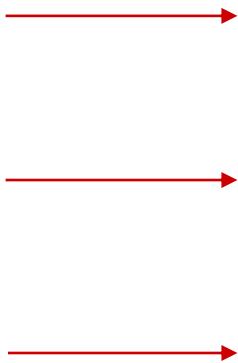
US National Research Council:
**“Photons are the new
Electrons.”**

Enabler

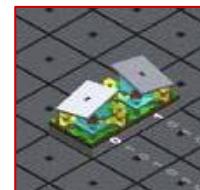
Novel Reconfigurable Optical Switches

→ **Spectrum** of prototypes

- Different sizes, different reconfiguration times
- From our ACM **SIGCOMM** workshop OptSys



Prototype 1



Prototype 2



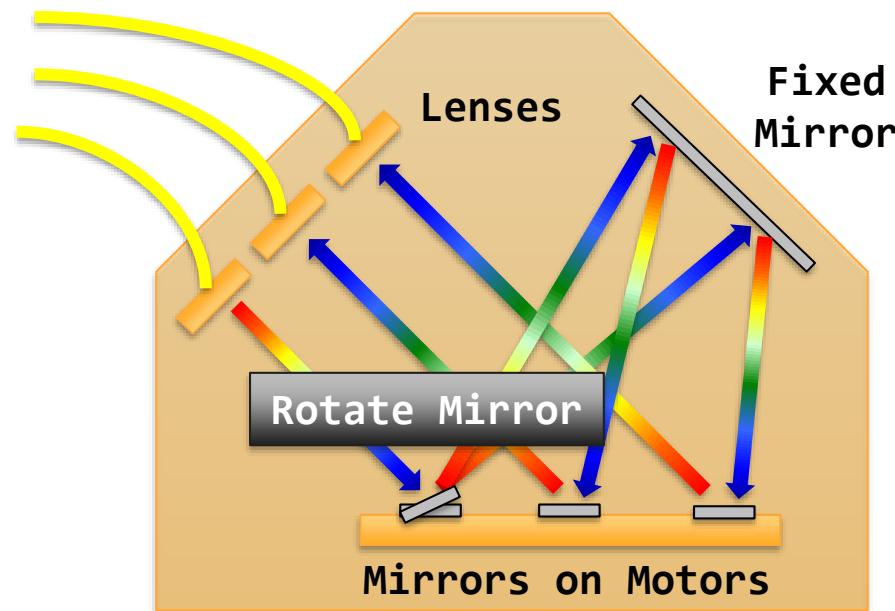
Prototype 3

Changing lambdas (ns)

Example

Optical Circuit Switch

- Optical Circuit Switch rapid adaption of physical layer
 - Based on rotating mirrors

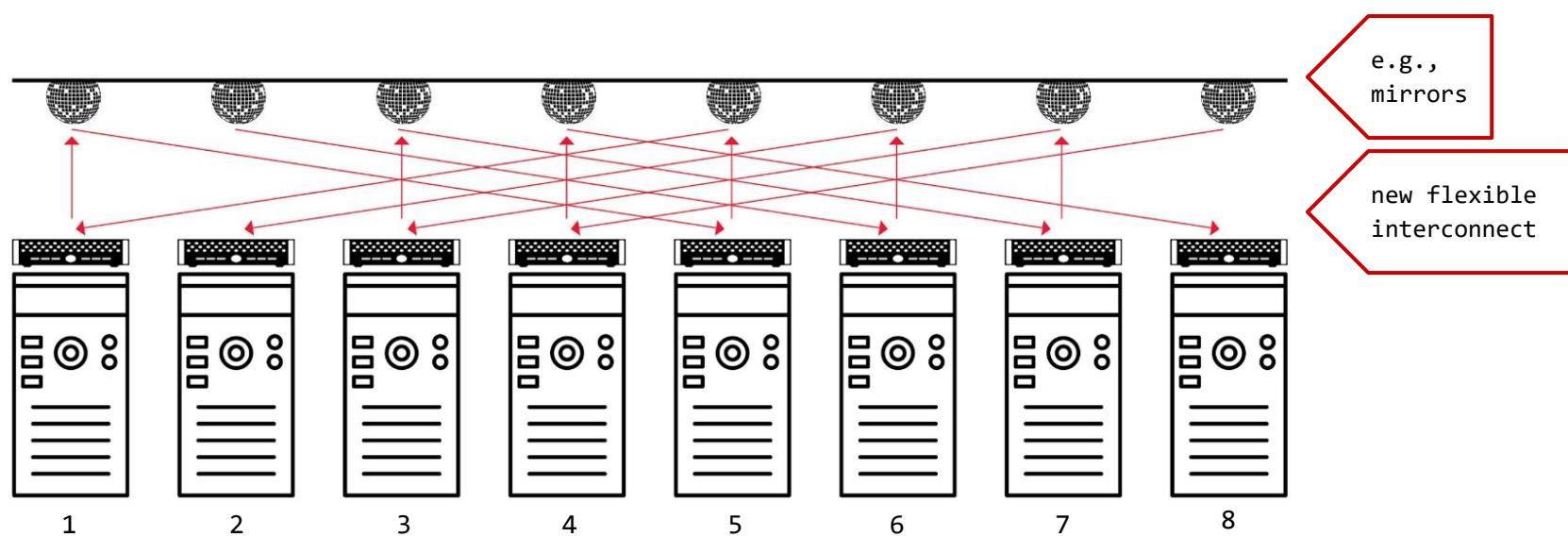


Optical Circuit Switch

By Nathan Farrington, SIGCOMM 2010

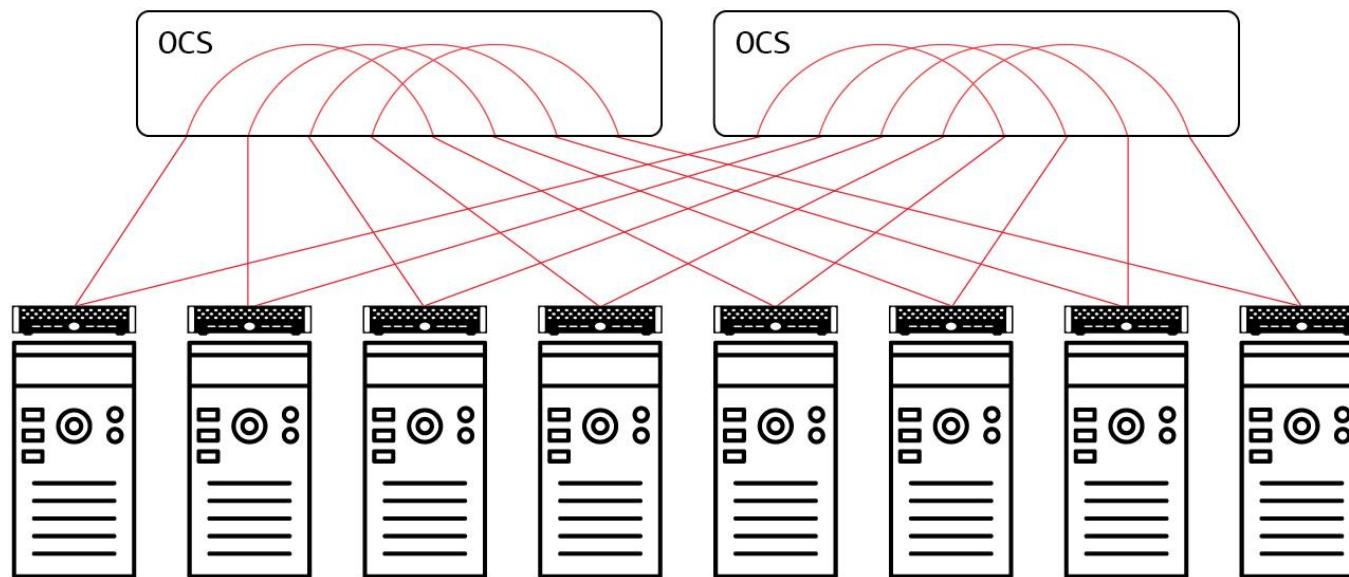
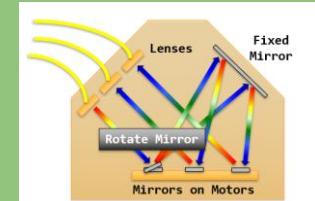
Recall: Our Vision

Using Mirrors and Lasers



Realization

with Optical Circuit Switches (OCS)



First Deployments

E.g., Google

Systems

Jupiter evolving: Reflecting on Google's data center network transformation

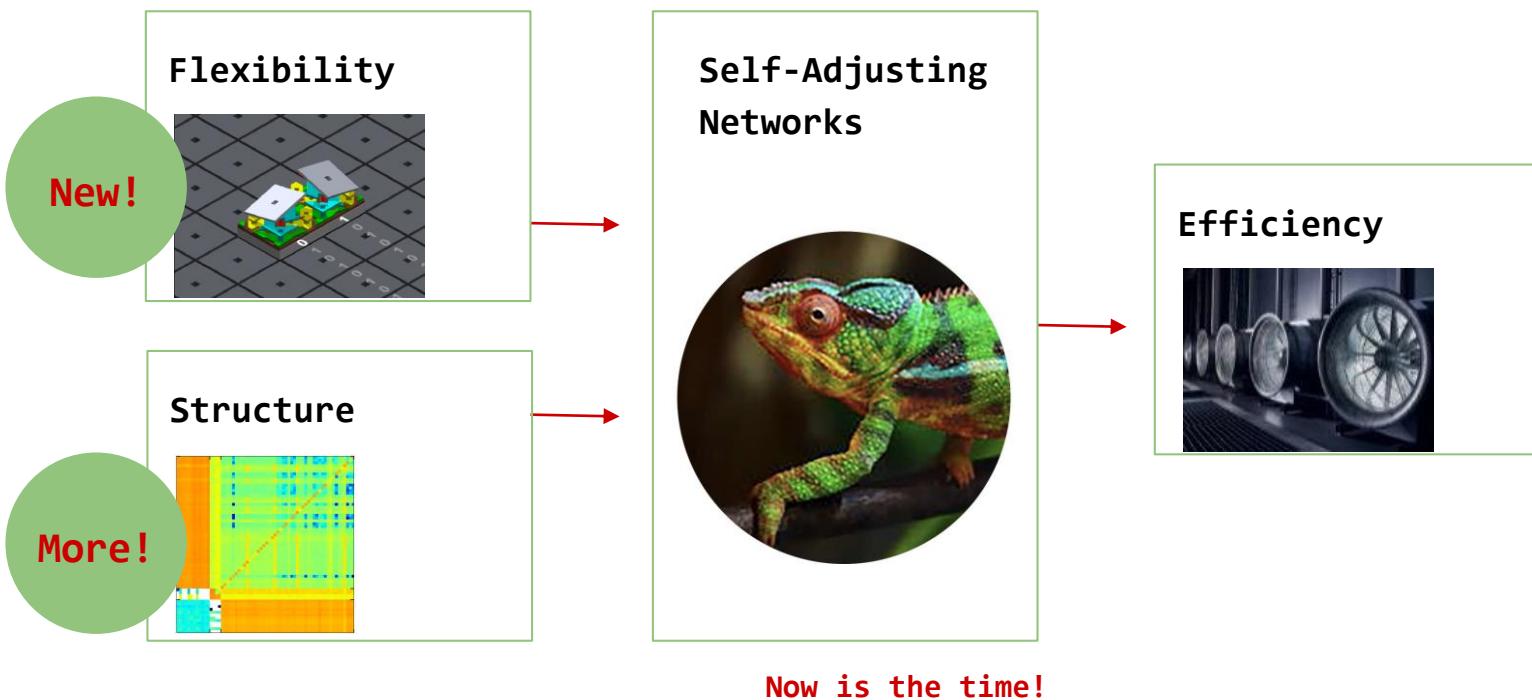
August 24, 2022

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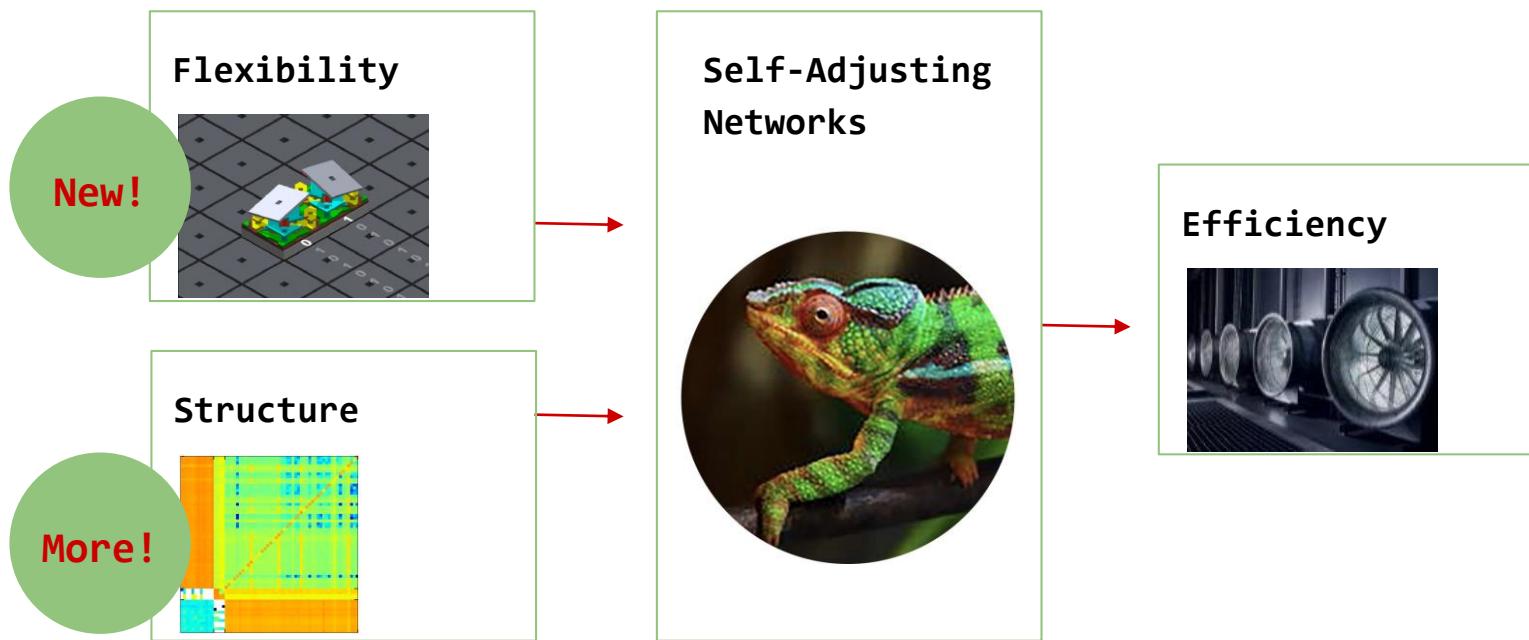
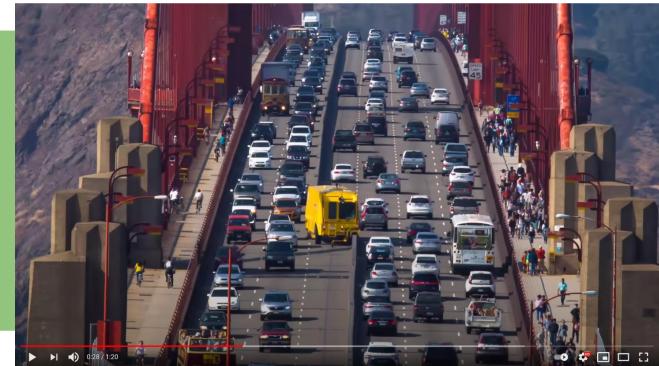
Amin Vahdat
VP & GM, Systems and Services Infrastructure

The Big Picture



The Big Picture

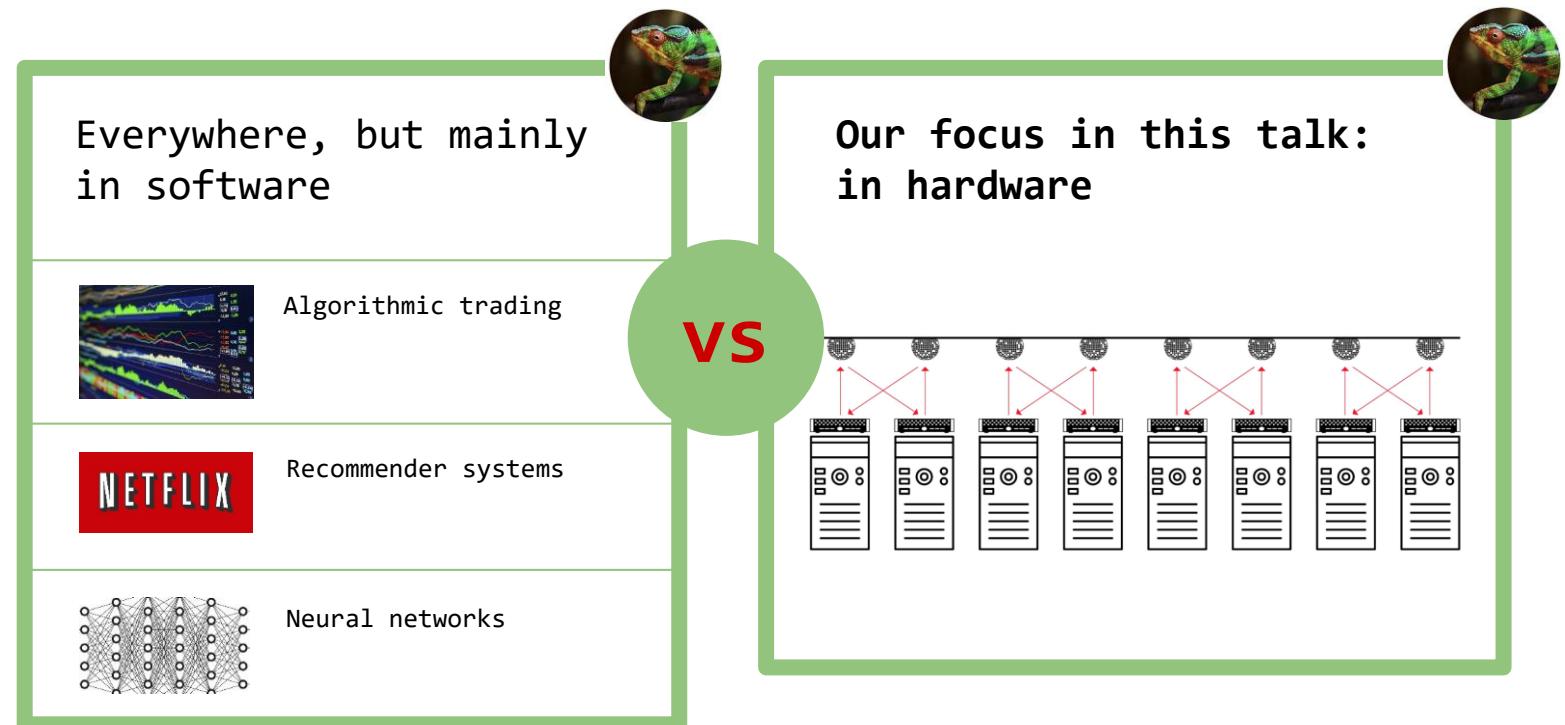
Like “Golden Gate Zipper”
for datacenters.



Now is the time!

Unique Position

Demand-Aware, Self-Adjusting Systems



First basic question:

How to measure and model
structure in workloads?

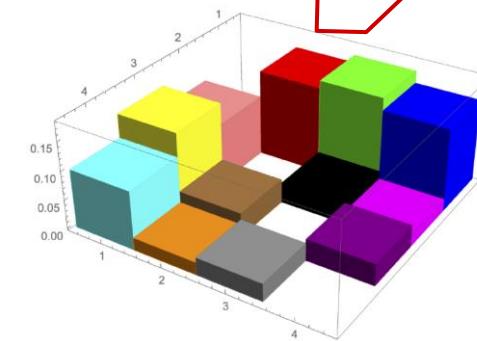
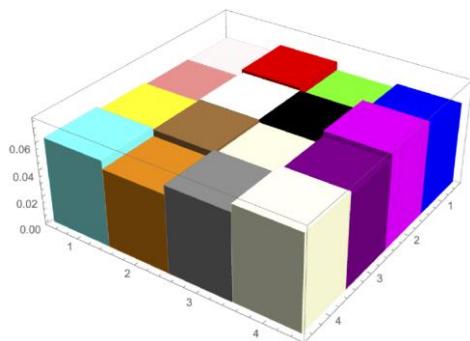
A first insight: related to entropy.

Intuition

Which demand has more structure?

→ Traffic matrices of two different distributed
ML applications

→ GPU-to-GPU



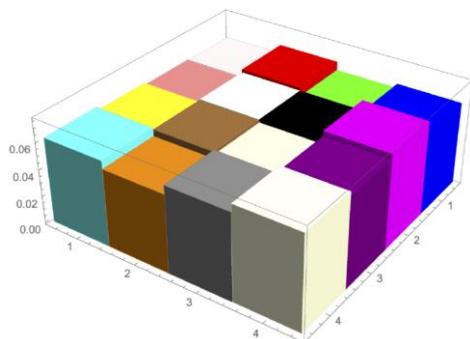
Color = communication pair

Intuition

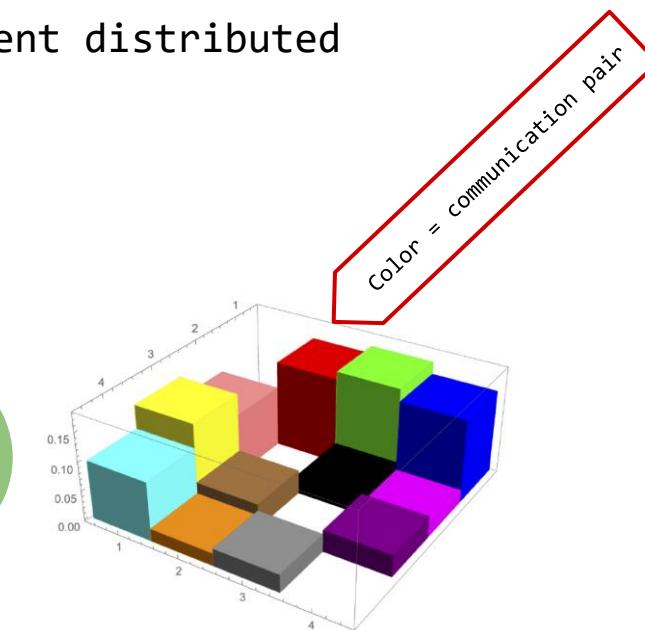
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More uniform

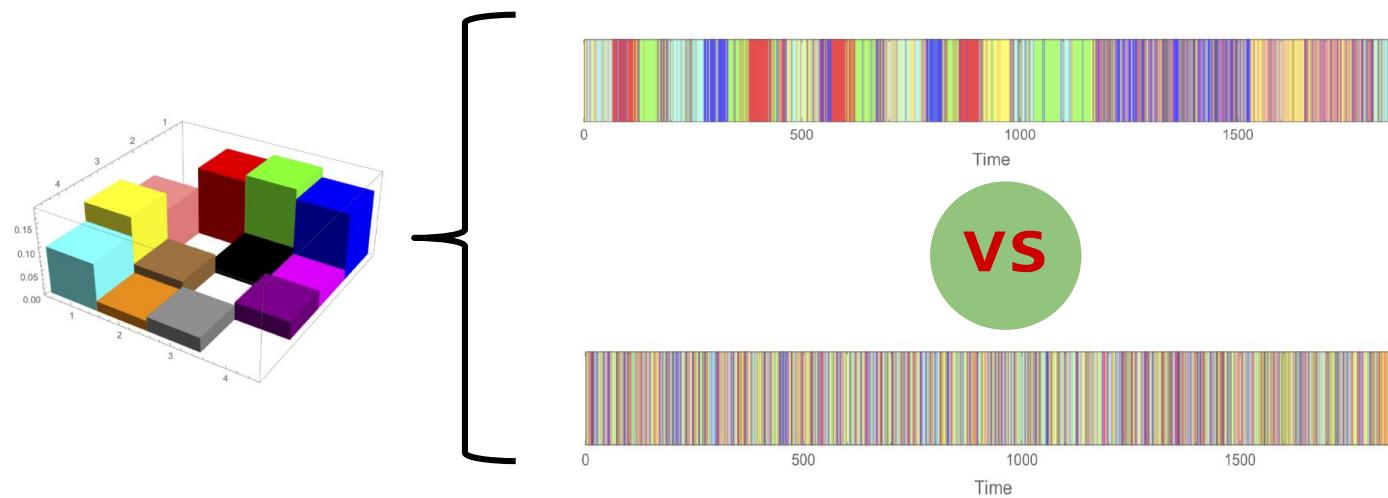


More structure

Intuition

Spatial vs temporal structure

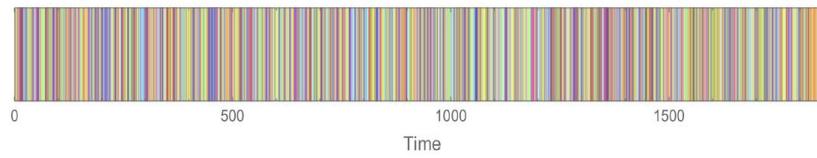
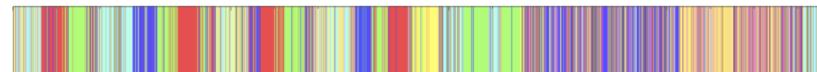
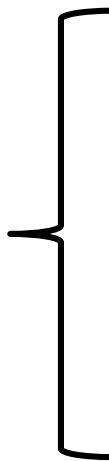
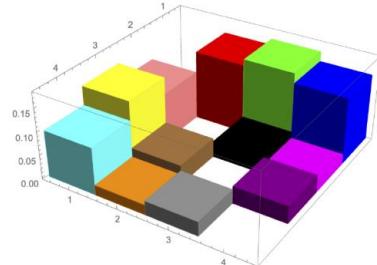
- Two different ways to generate same traffic matrix:
 - Same non-temporal structure
- Which one has more structure?



Intuition

Spatial vs temporal structure

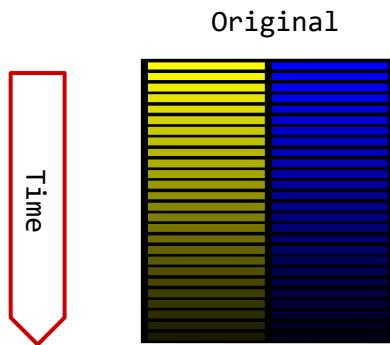
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Systematically?

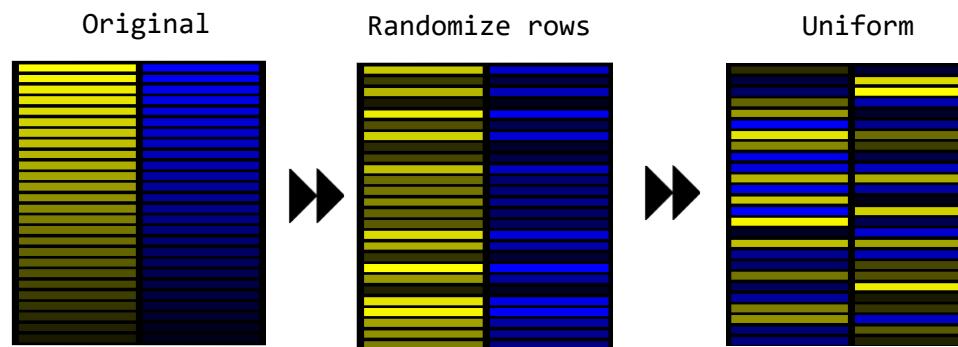
Trace Complexity

Information-Theoretic Approach
“Shuffle&Compress”



Trace Complexity

Information-Theoretic Approach
“Shuffle&Compress”

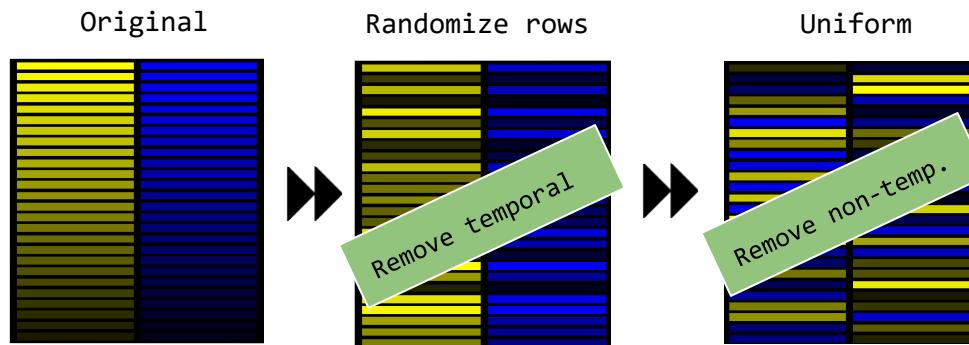


Increasing complexity (systematically randomized)

More structure (compresses better)

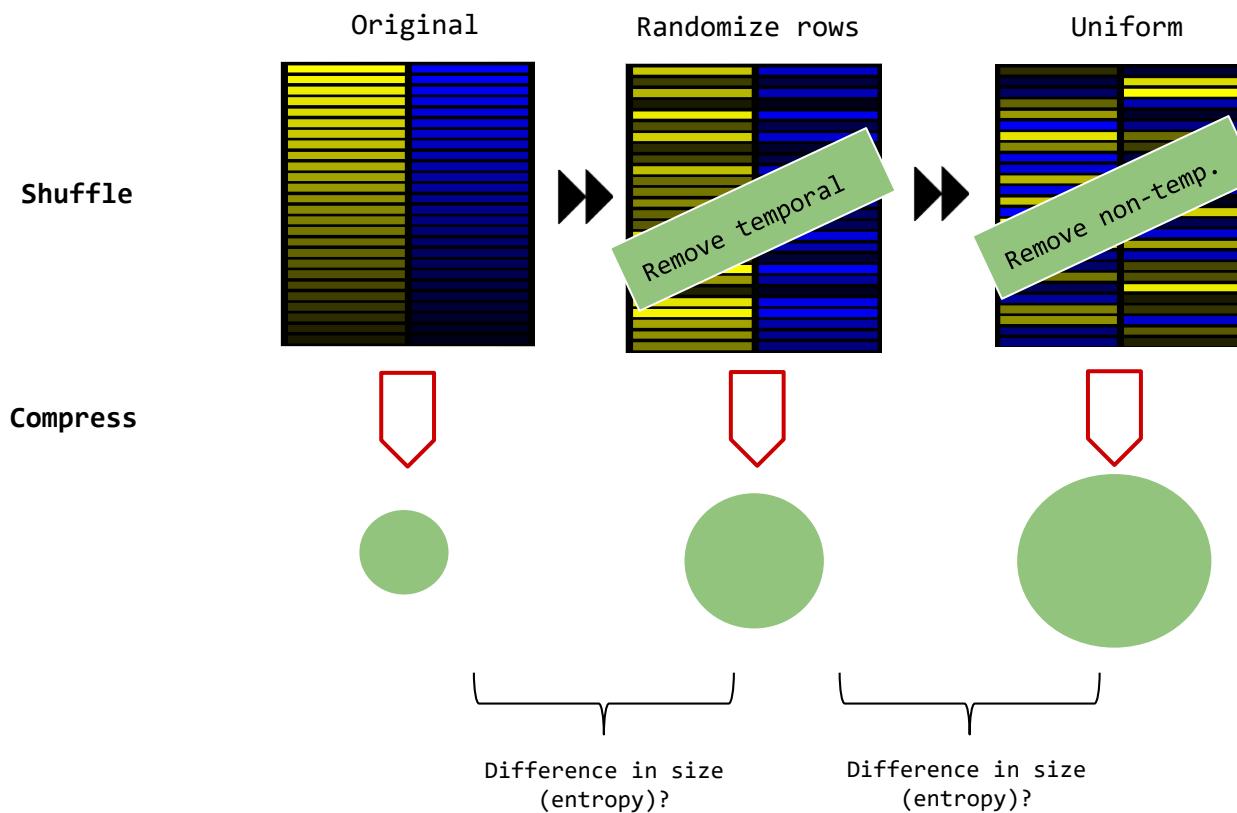
Trace Complexity

Information-Theoretic Approach
“Shuffle&Compress”



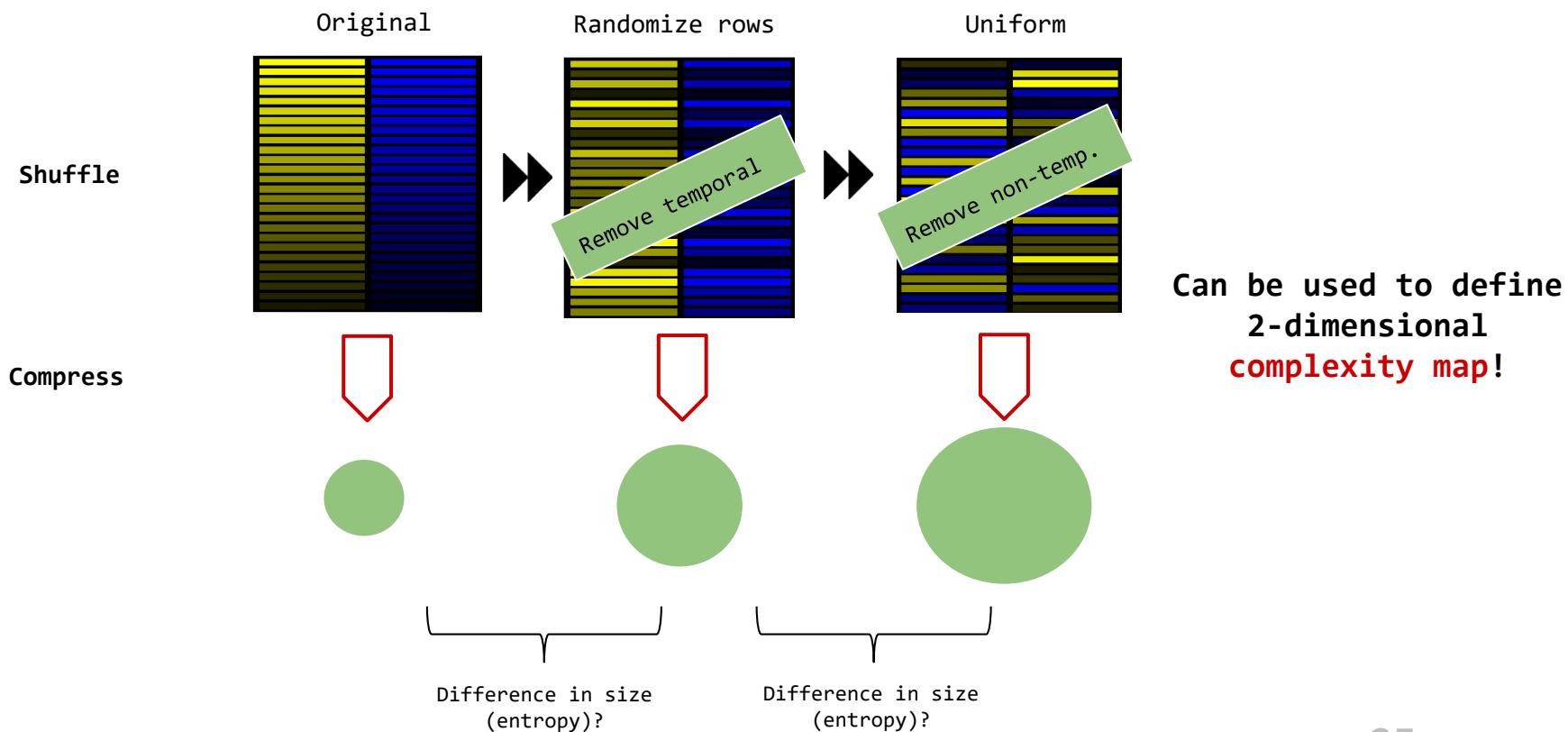
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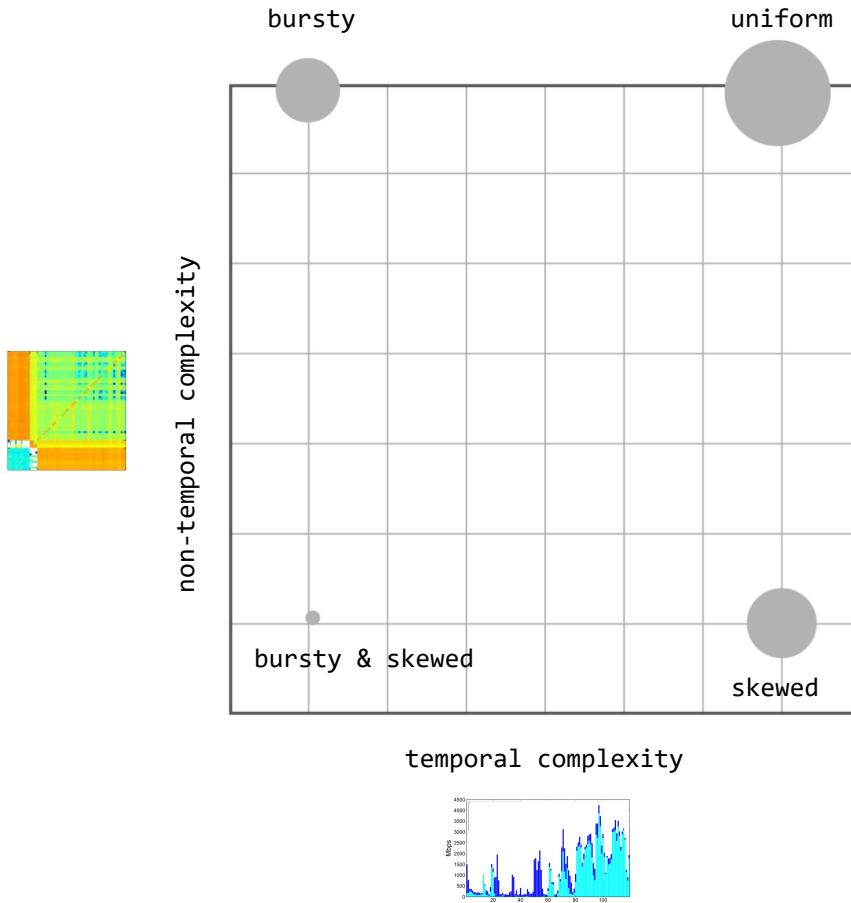


Trace Complexity

Information-Theoretic Approach
“Shuffle&Compress”



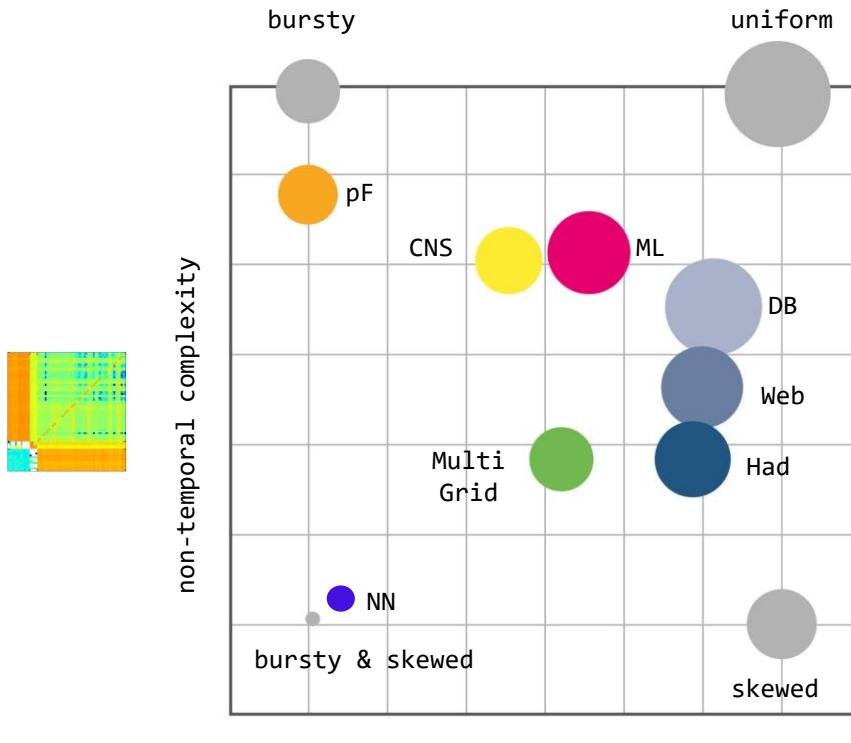
Complexity Map



No structure

Our approach: iterative
randomization and
compression of trace to
identify dimensions of
structure.

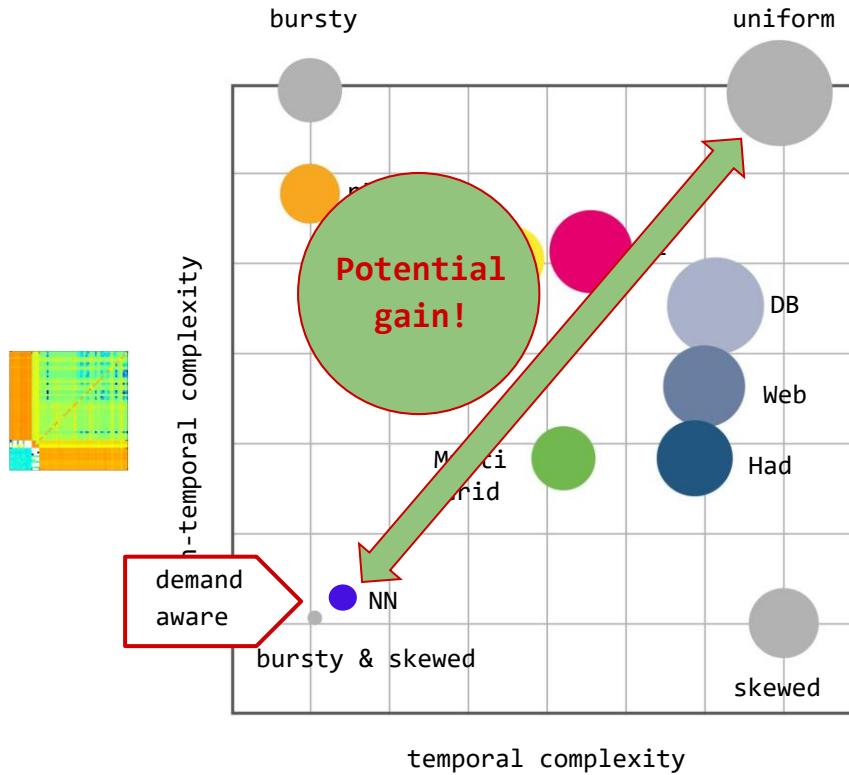
Complexity Map



Our approach: iterative randomization and compression of trace to identify dimensions of structure.

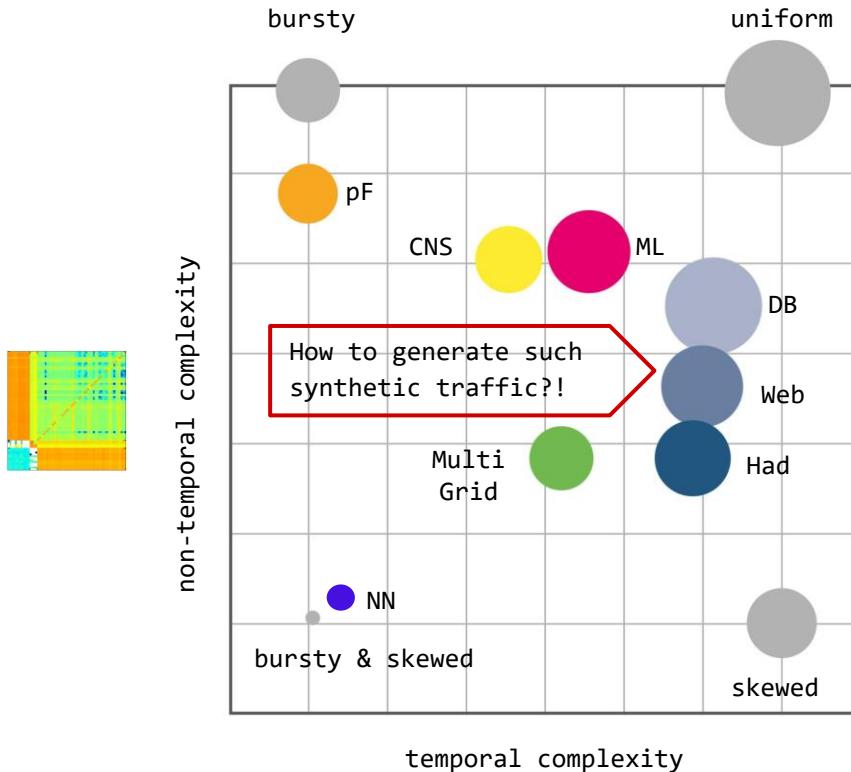
Different structures!

Complexity Map



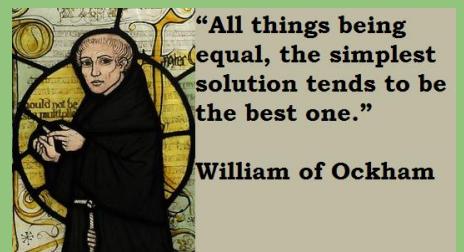
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Complexity Map

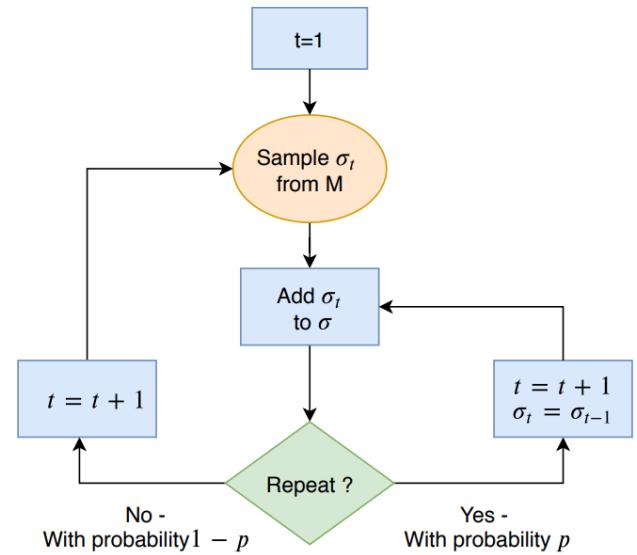


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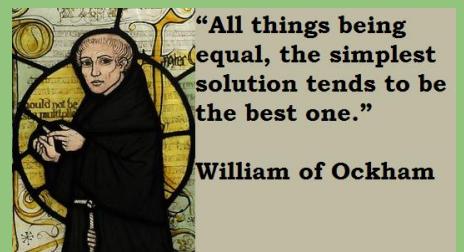
From Analysis to Synthesis



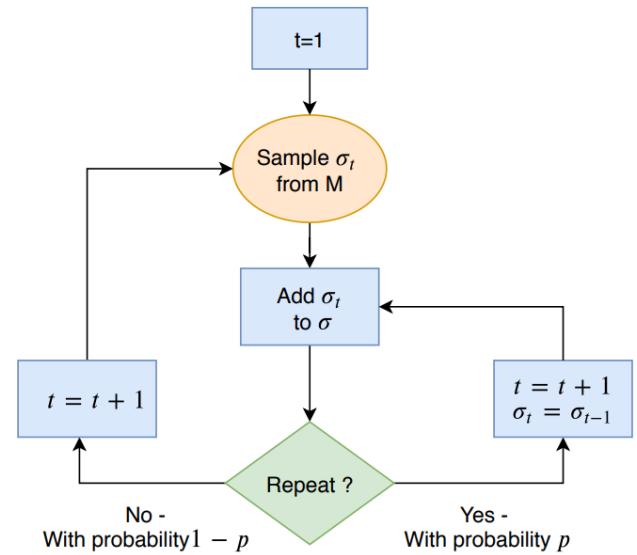
- Complexity map is just 2-dimensional: many ways to synthesize any point on map
- Most simple ("Occam's razor"):
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Further Reading

On the Complexity of Traffic Traces and Implications

Chen Avin, Manya Ghobadi, Chen Griner, and Stefan Schmid.

ACM SIGMETRICS and ACM Performance Evaluation Review (PER), Boston, Massachusetts, USA, June 2020.

Analyzing the Communication Clusters in Datacenters

Klaus-Tycho Foerster, Thibault Marette, Stefan Neumann, Claudia Plant, Ylli Sadikaj, Stefan Schmid, and Yllka Velaj.

The Web Conference (WWW), Austin, Texas, USA, April 2023.

Network Traffic Characteristics of Machine Learning Frameworks Under the Microscope

Johannes Zerwas, Kaan Aykurt, Stefan Schmid, and Andreas Blenk. 17th International Conference on Network and Service Management (CNSM), Izmir, Turkey, October 2021.

Website: trace-collection.net



The Natural Question:

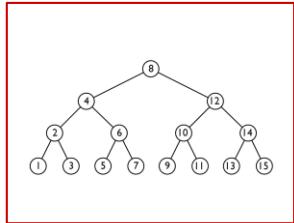
Given This Structure,
What Can Be Achieved?
Metrics and Algorithms?

Also depends on entropy of the demand!

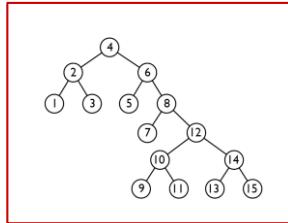
Insight:

Connection to Datastructures

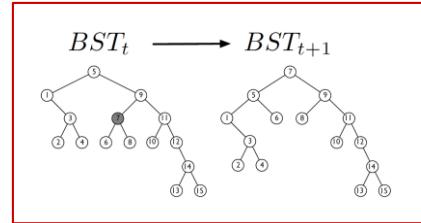
Traditional BST



Demand-aware BST



Self-adjusting BST

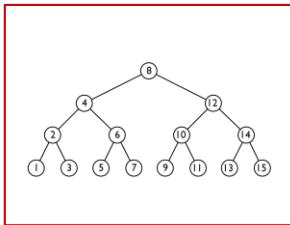


More structure: improved **access cost**

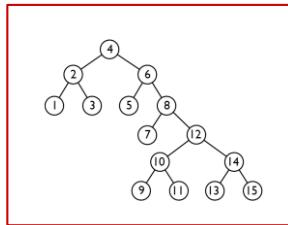
Insight:

Connection to Datastructures & Coding

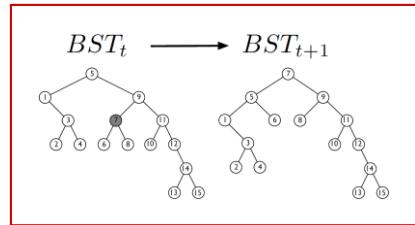
Traditional BST
(Worst-case coding)



Demand-aware BST
(Huffman coding)



Self-adjusting BST
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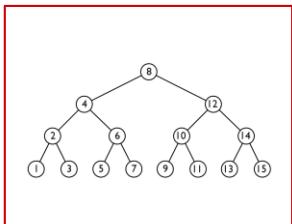


More structure: improved **access cost** / shorter **codes**

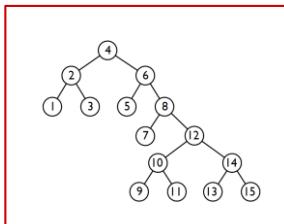
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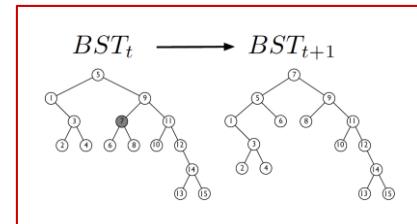
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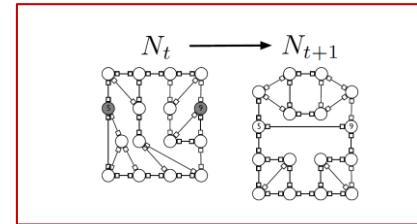
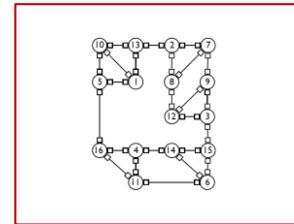
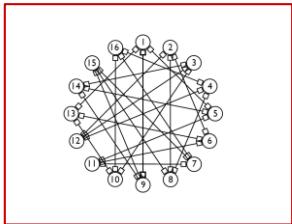
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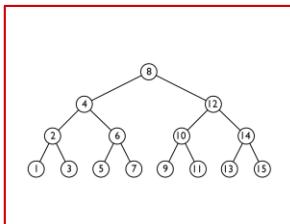


Similar **benefits?**

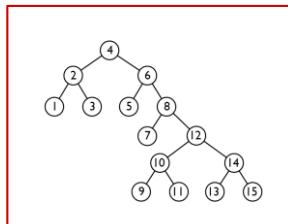
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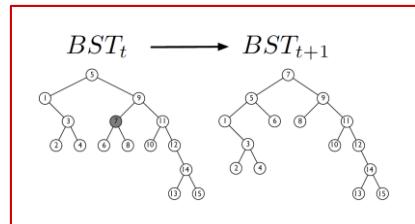
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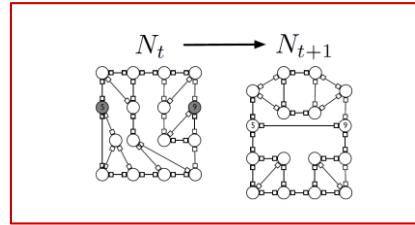
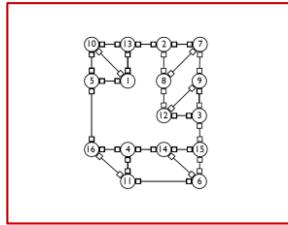
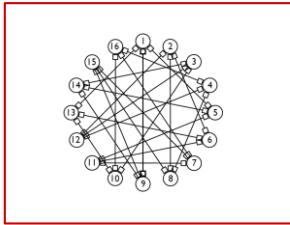


Self-adjusting BST
(Dynamic Huffman coding)



More than
an analogy!

More structure: improved **access cost** / shorter **codes**

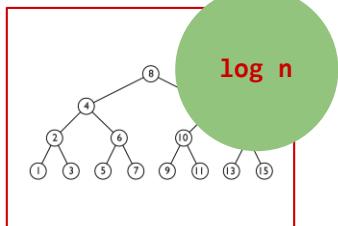


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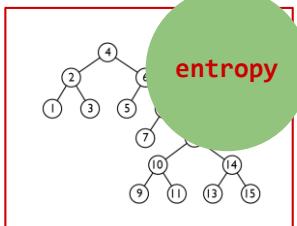
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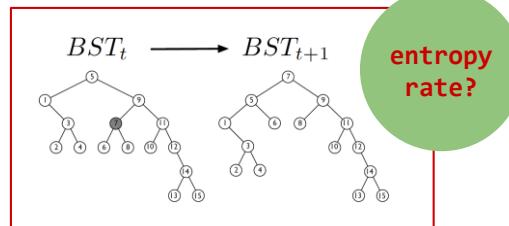
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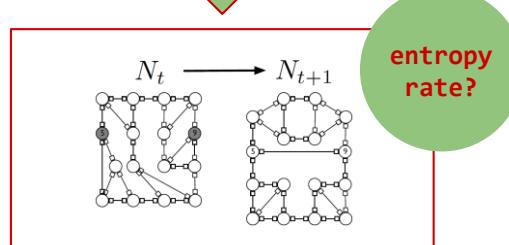
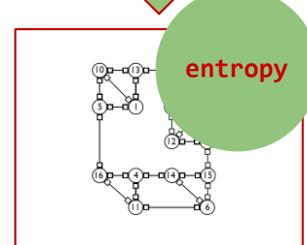
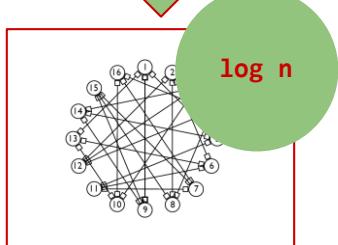
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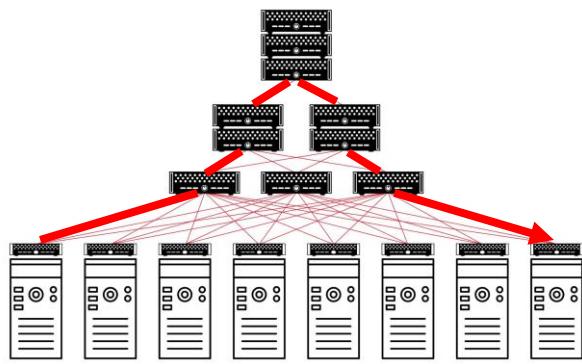
Reduced expected **route lengths**!

Generalize methodology:
... and transfer
entropy bounds and
algorithms of data-
structures to networks.

First result:
Demand-aware networks
of asymptotically
optimal route lengths.

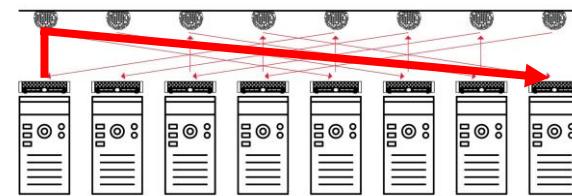
Reality more complicated

- Self-adjusting networks may be really useful to serve large flows (**elephant flows**): avoiding multi-hop routing



6 hops

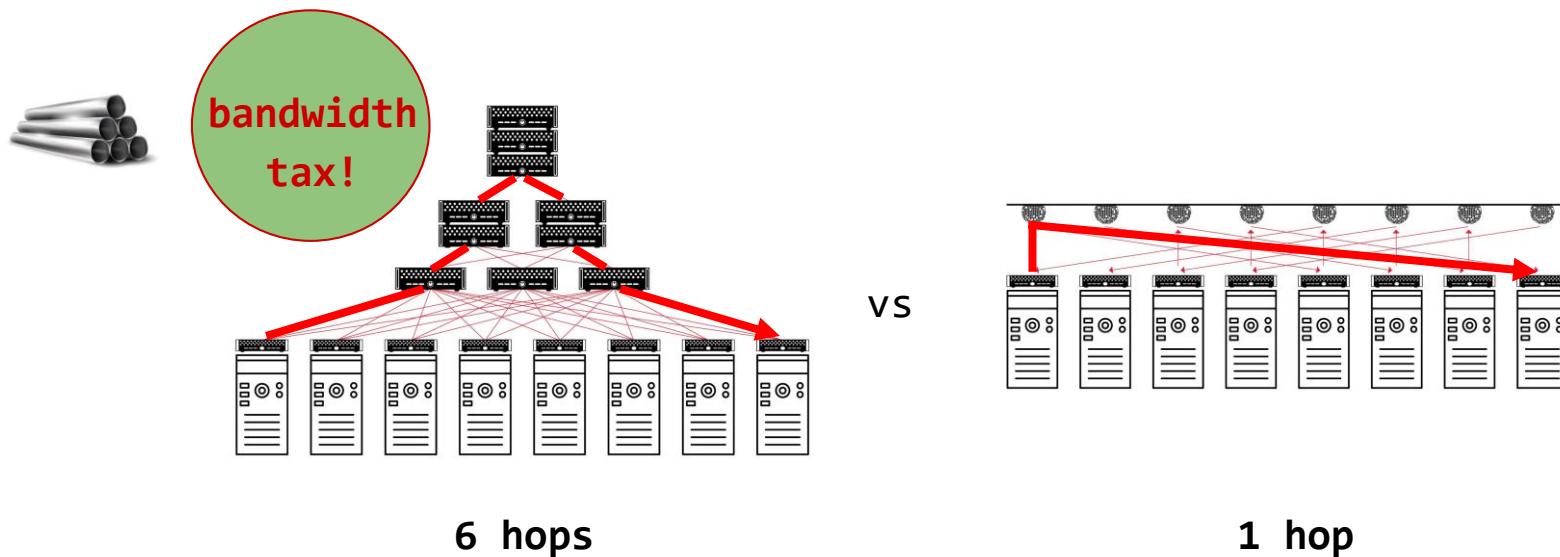
vs



1 hop

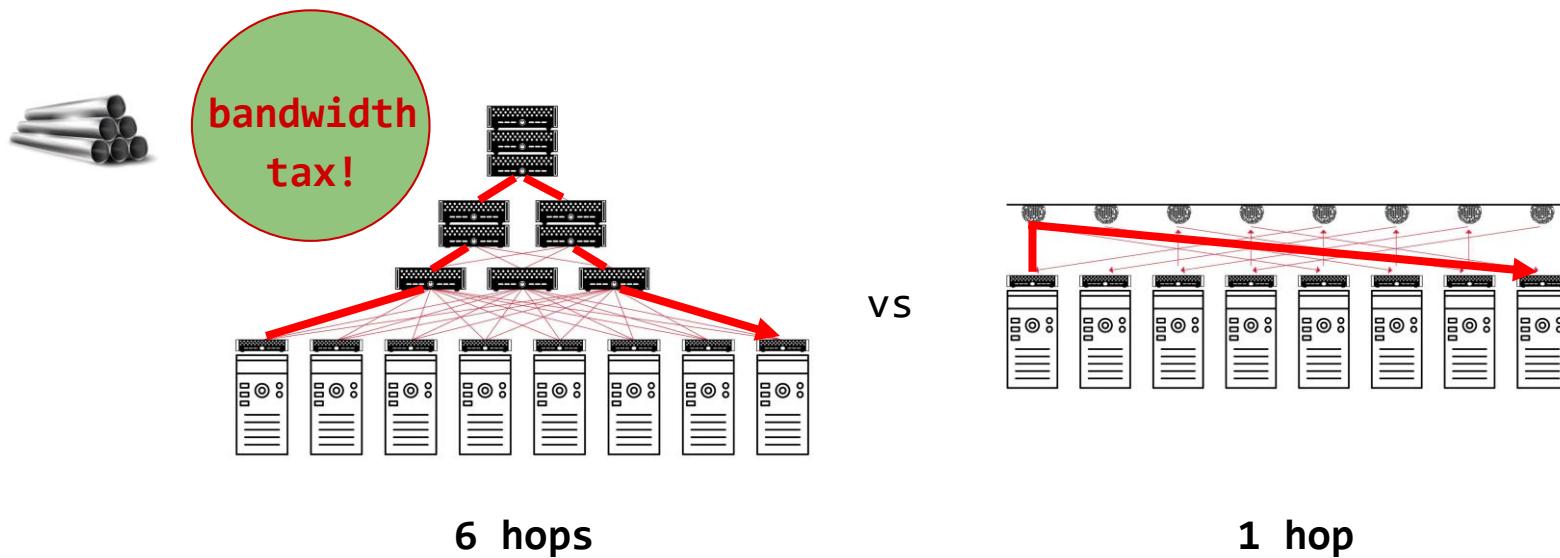
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Reality more complicated

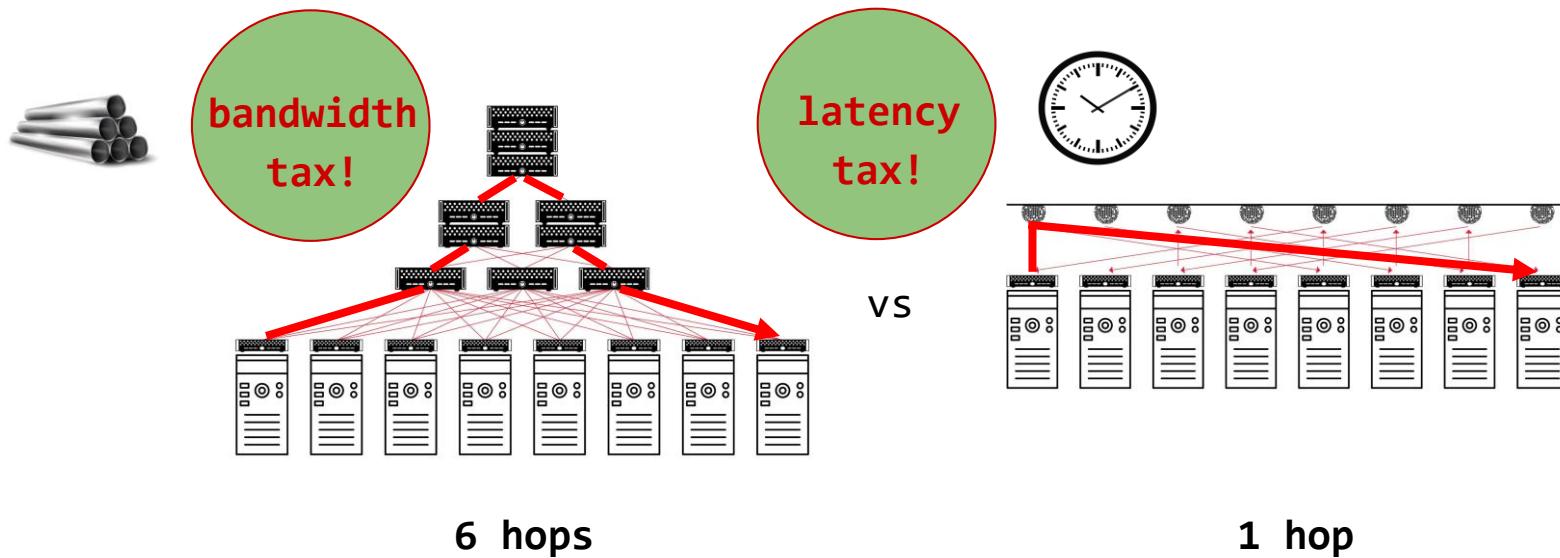
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- However, requires optimization and adaption, which **takes time**

Reality more complicated

- Self-adjusting networks may be really useful to serve large flows (**elephant flows**): avoiding multi-hop routing



- However, requires optimization and adaption, which **takes time**

Indeed, it is more complicated than that...

Challenge: Traffic Diversity

Diverse patterns:

- Shuffling/Hadoop:
all-to-all
- All-reduce/ML: **ring** or
tree traffic patterns
 - **Elephant** flows
- Query traffic: skewed
 - **Mice** flows
- Control traffic: does not evolve
but has non-temporal structure

Diverse requirements:

- ML is **bandwidth** hungry,
small flows are **latency-**
sensitive



Shuffling
All-to-All



ML
Large flows



Delay
sensitive



Telemetry
/ control

Opportunity: Tech Diversity

Diverse topology components:

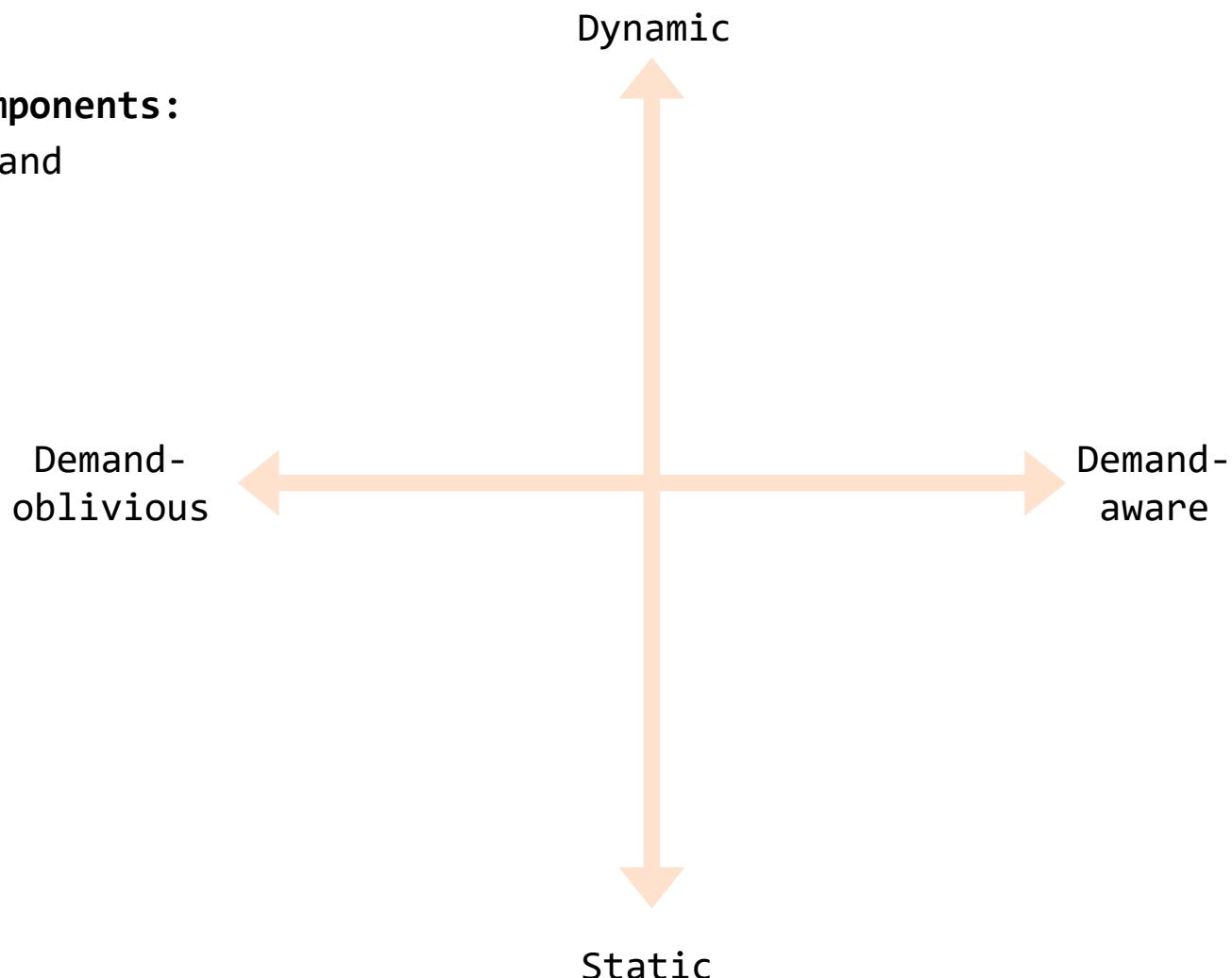
- demand-**oblivious** and
- demand-**aware**



Opportunity: Tech Diversity

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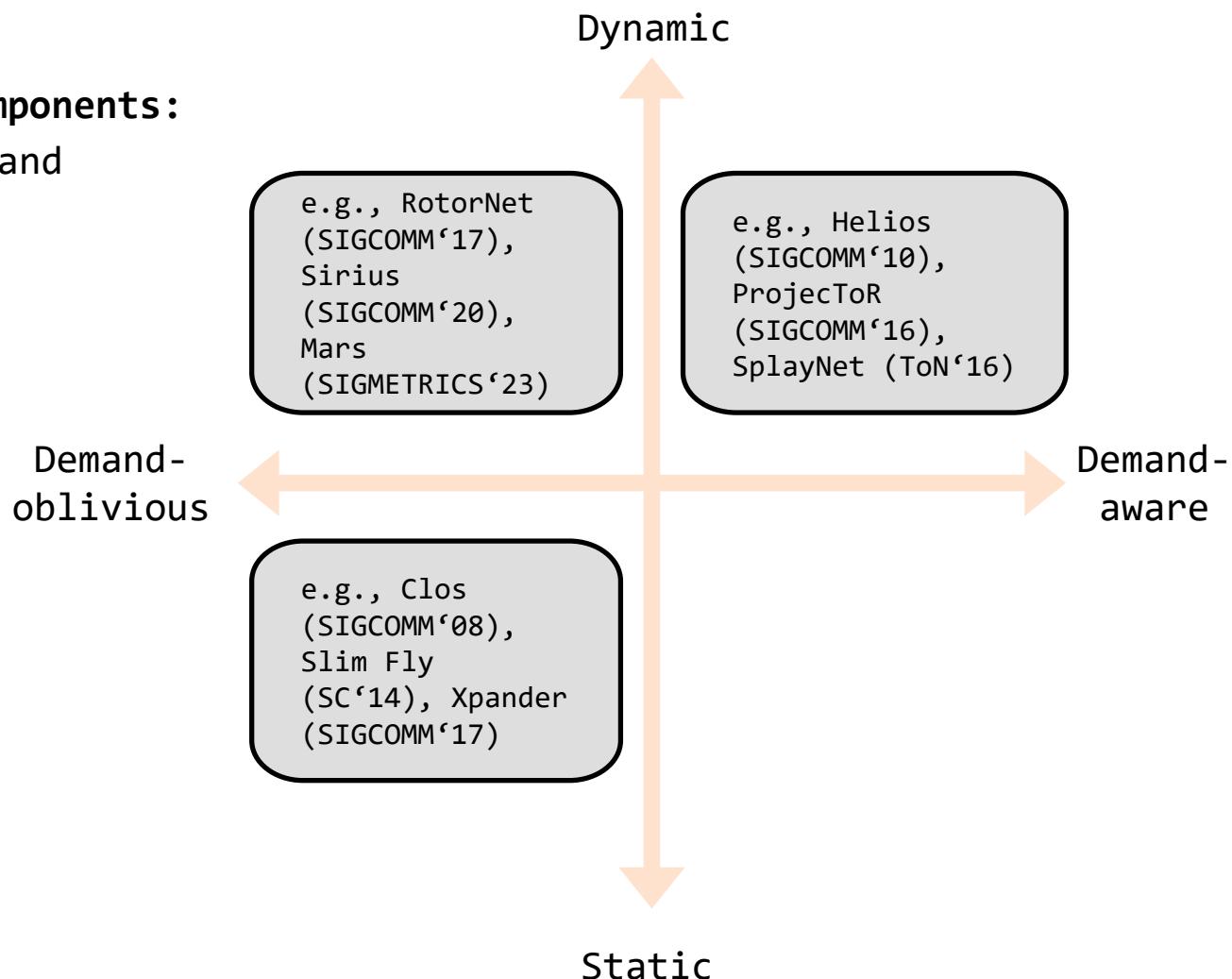
- demand-**oblivious** and
demand-**aware**
- static vs dynamic



Opportunity: Tech Diversity

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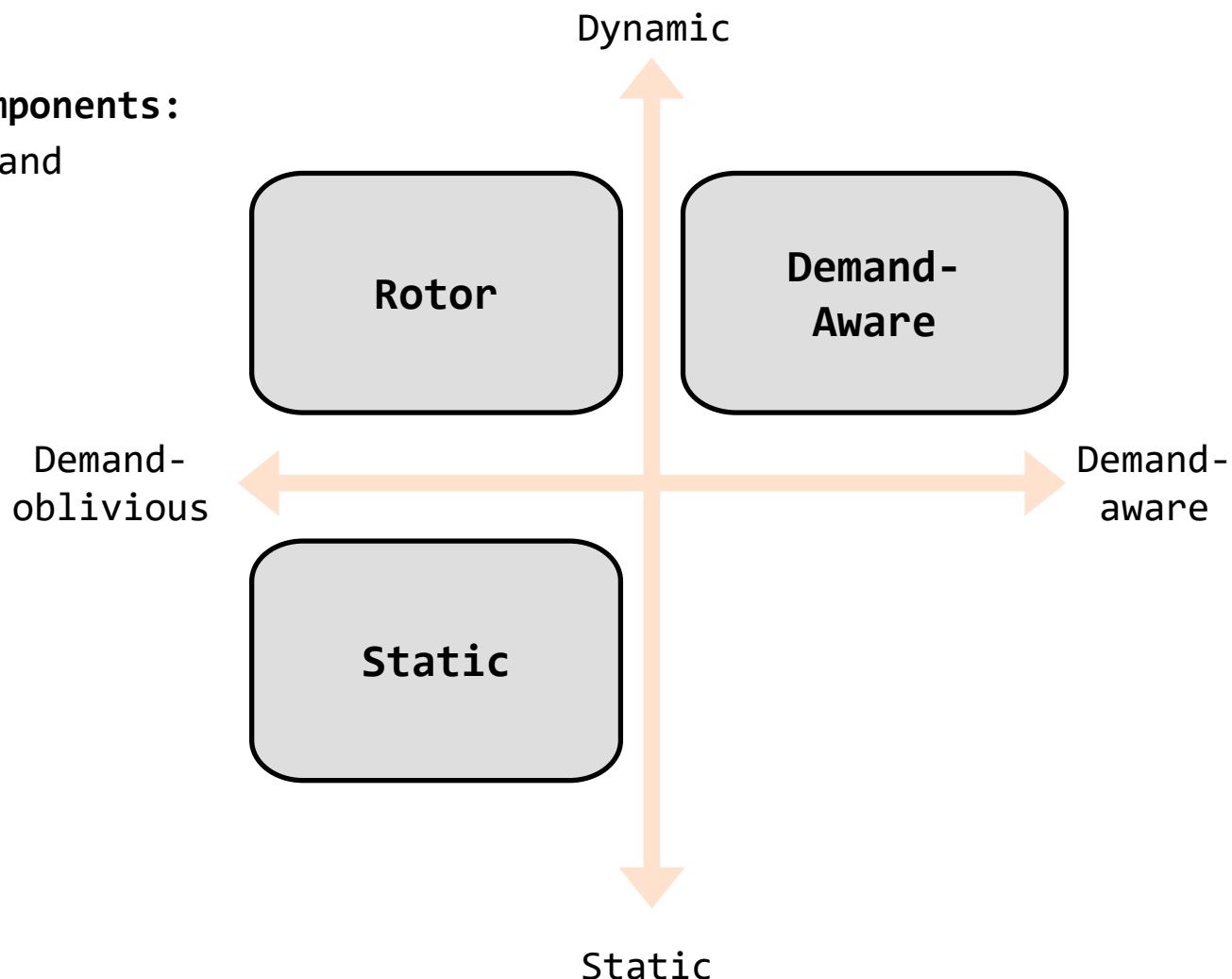
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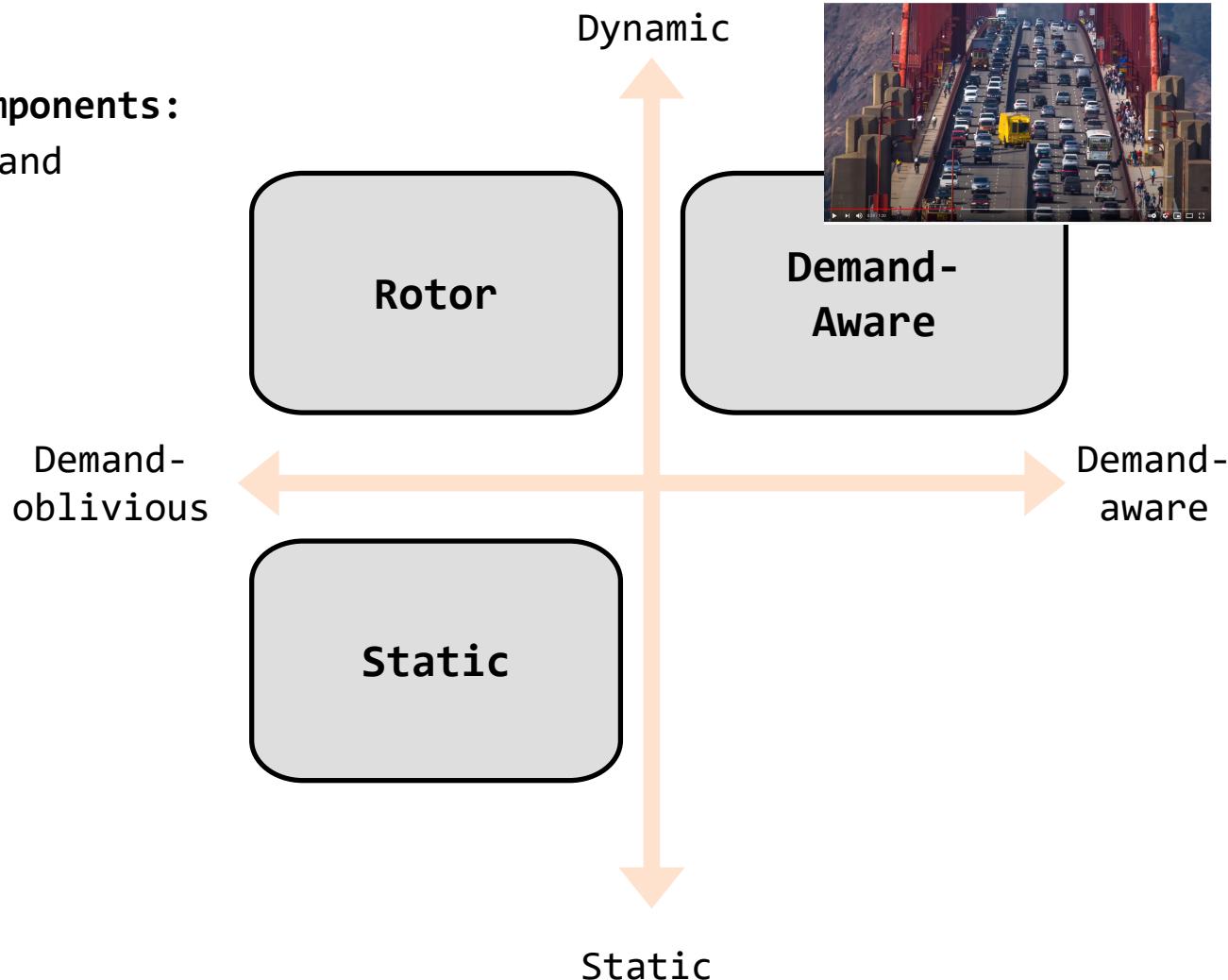
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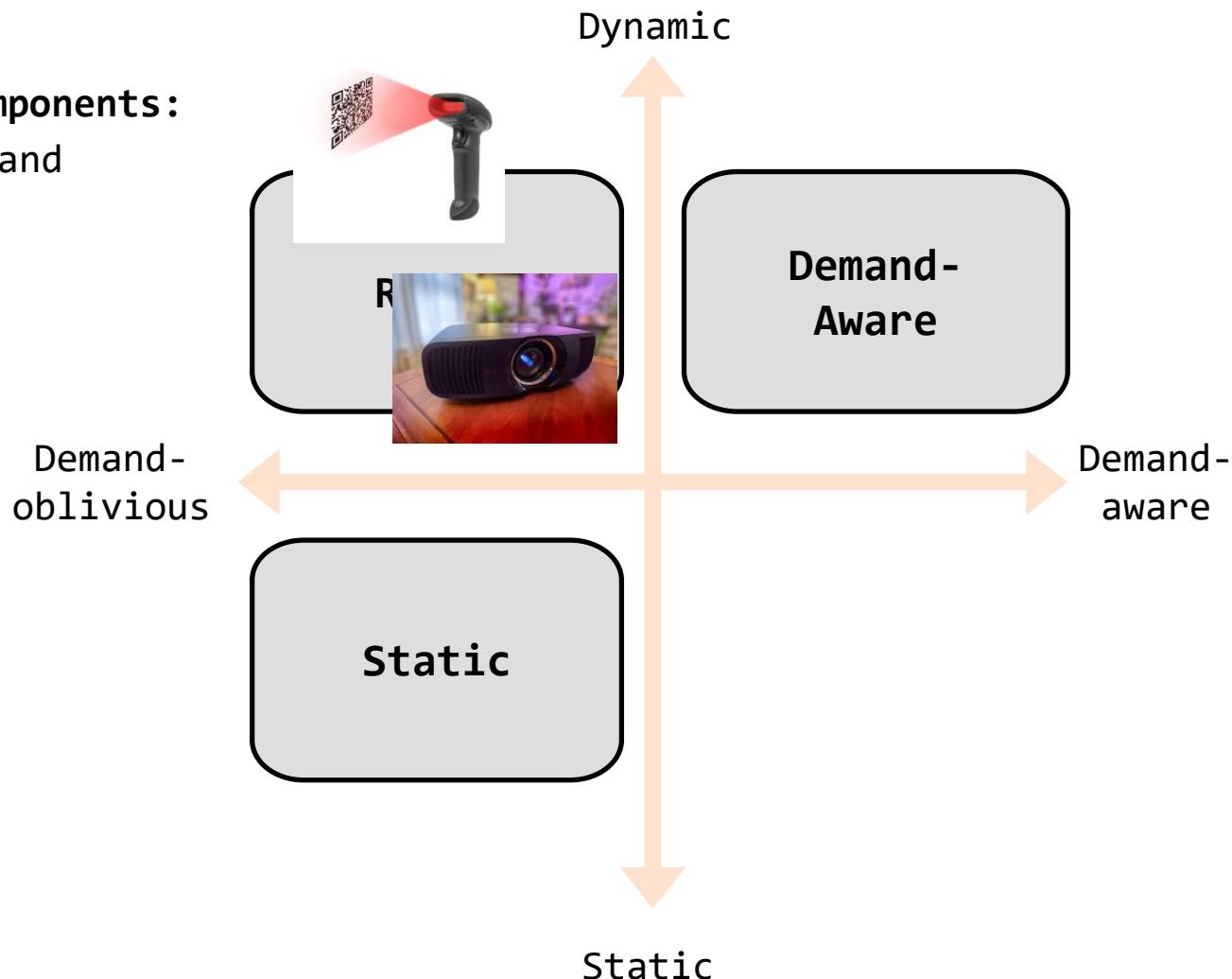
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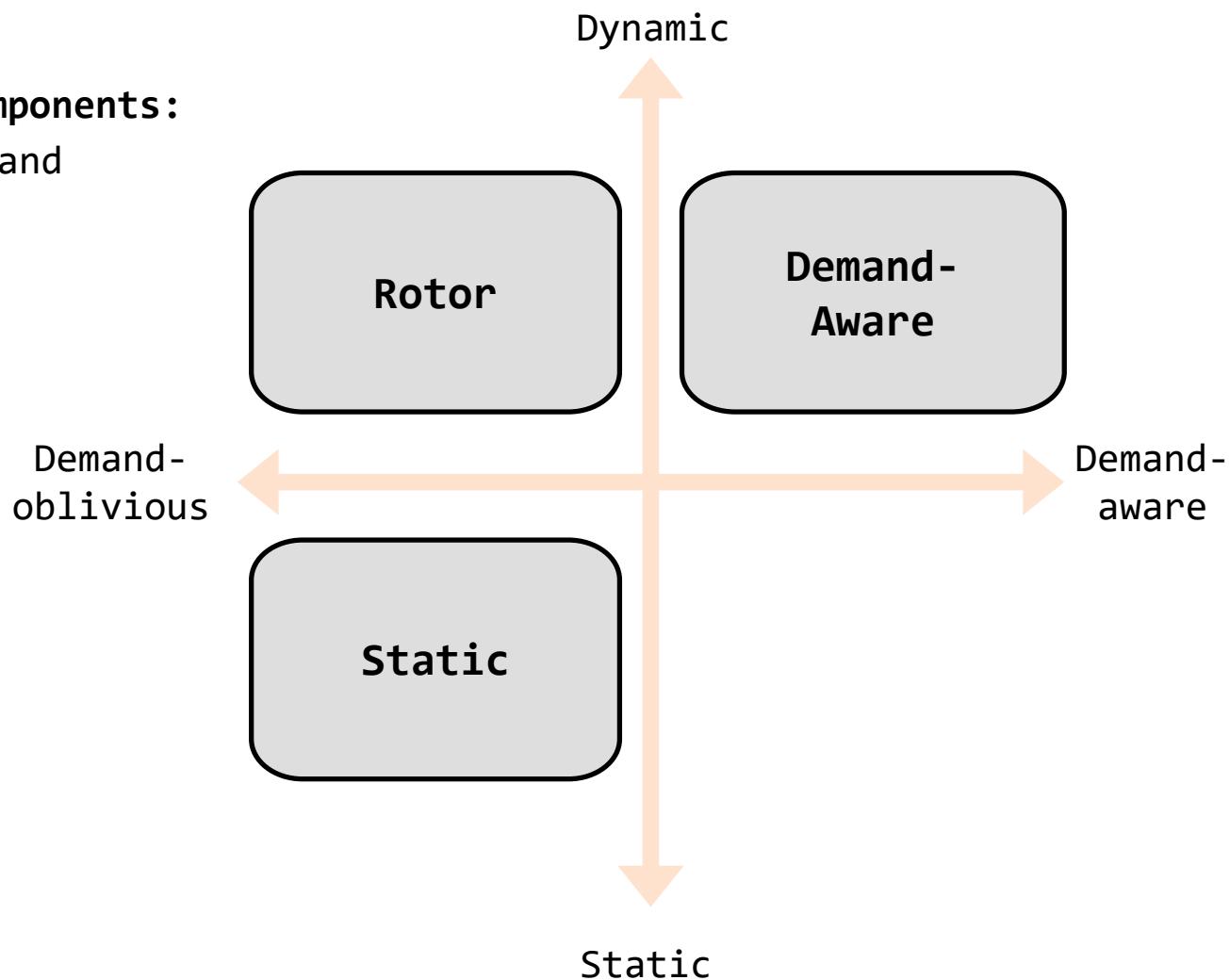
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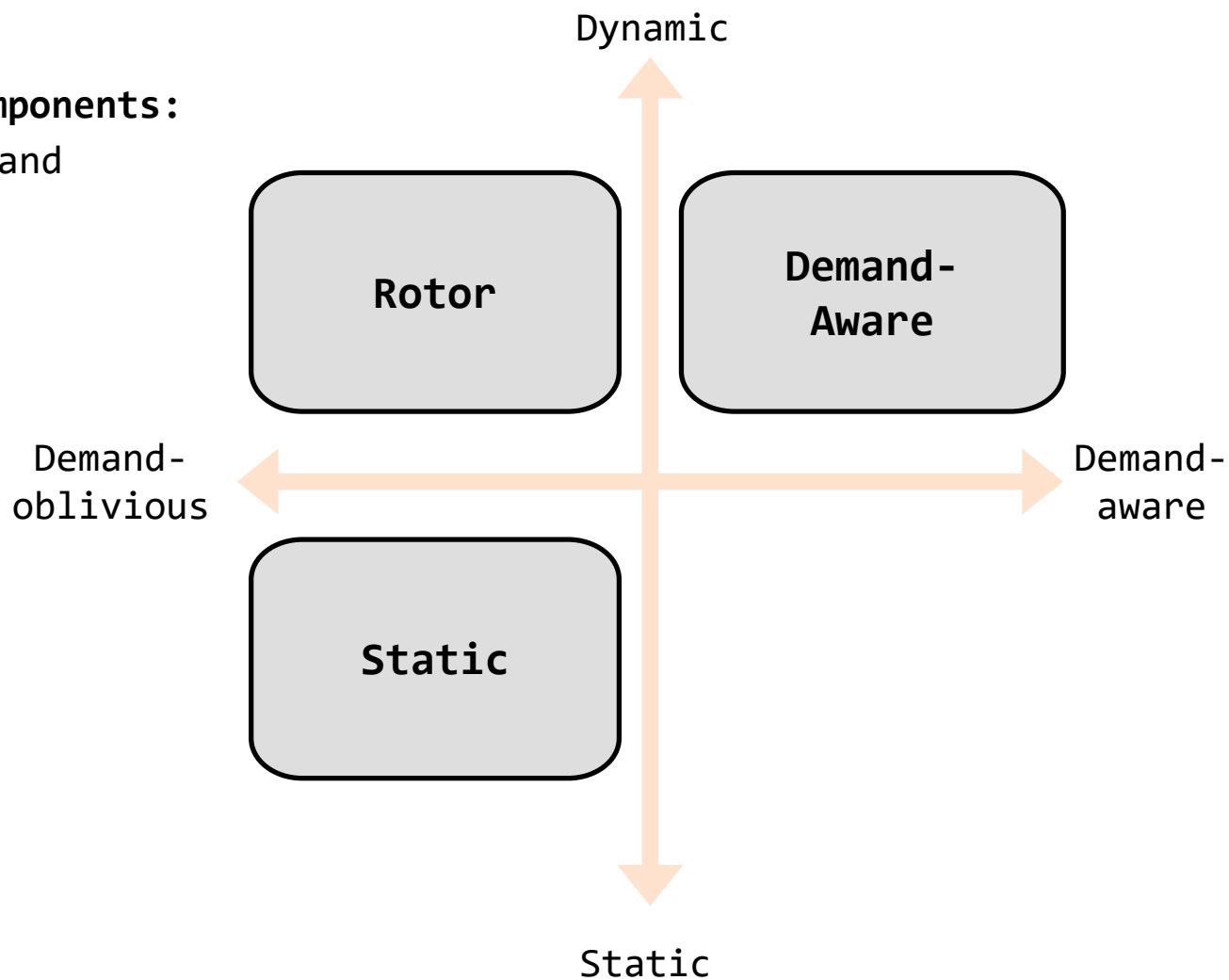
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Opportunity: Tech Diversity

Diverse topology components:

- demand-**oblivious** and
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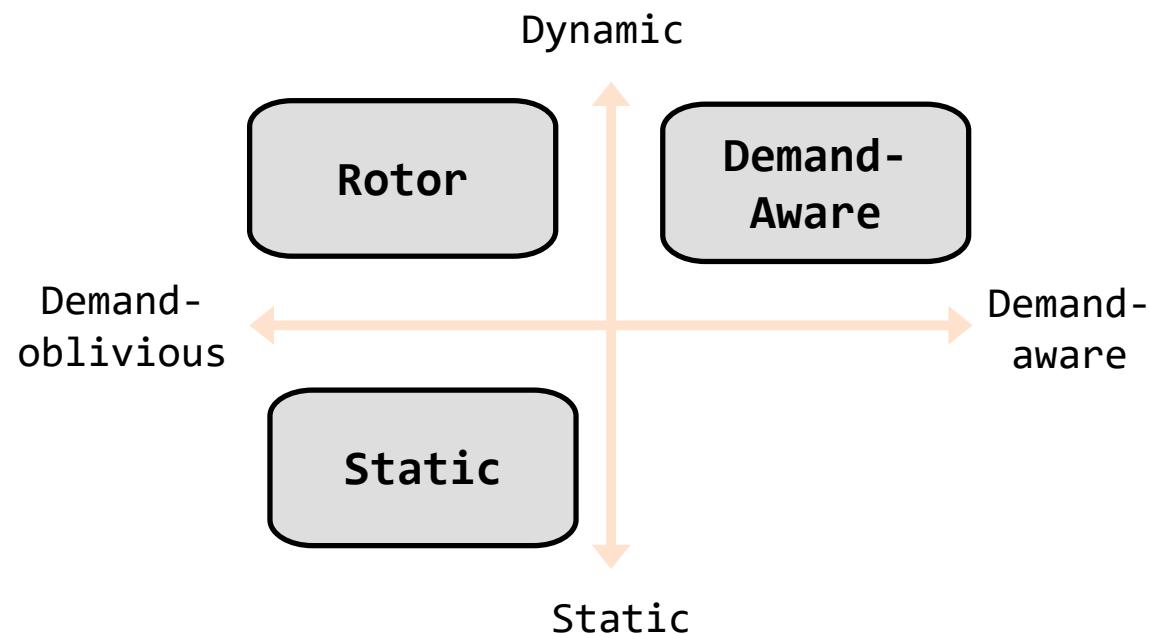
As always in CS:
It depends...

Examples:

Match or Mismatch?

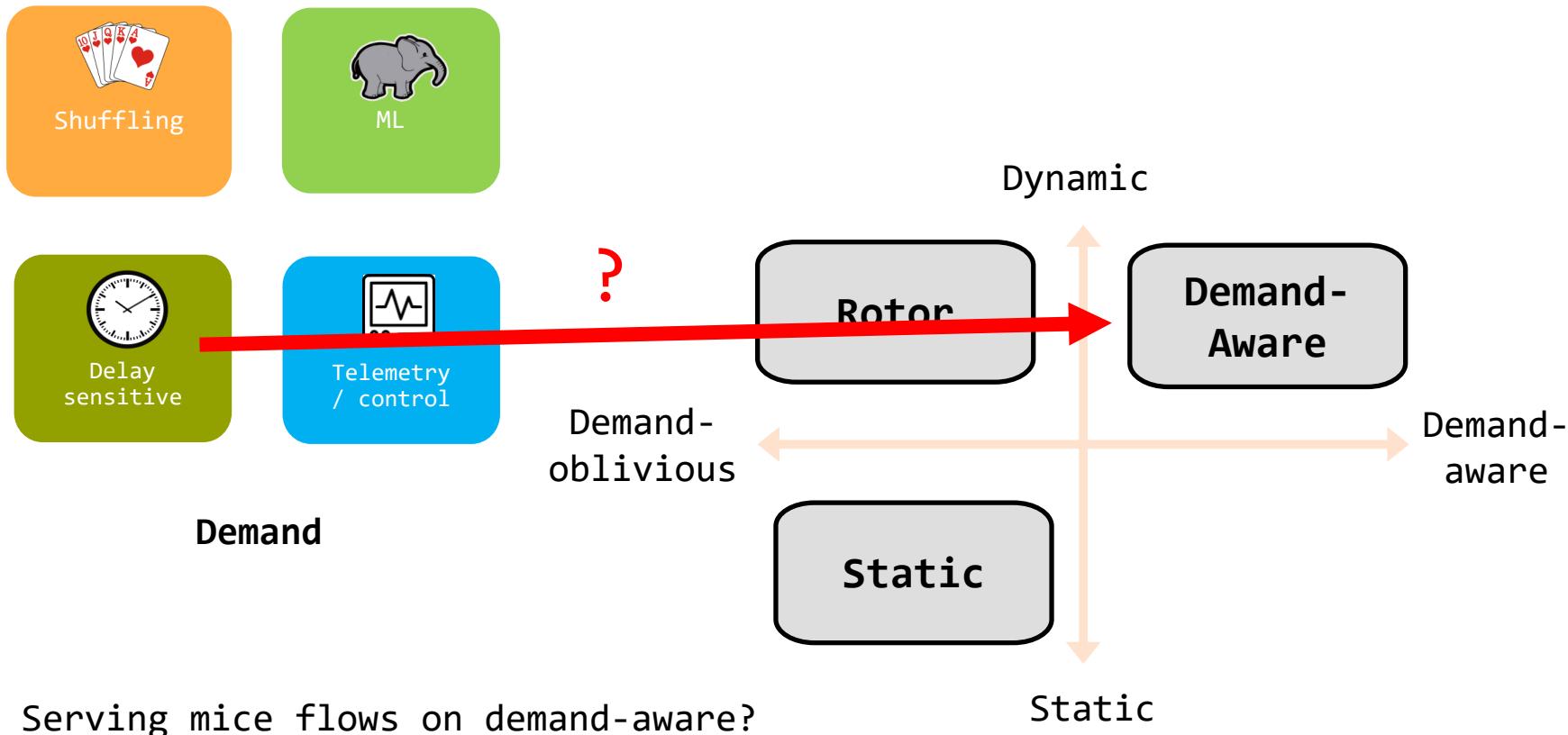


Demand



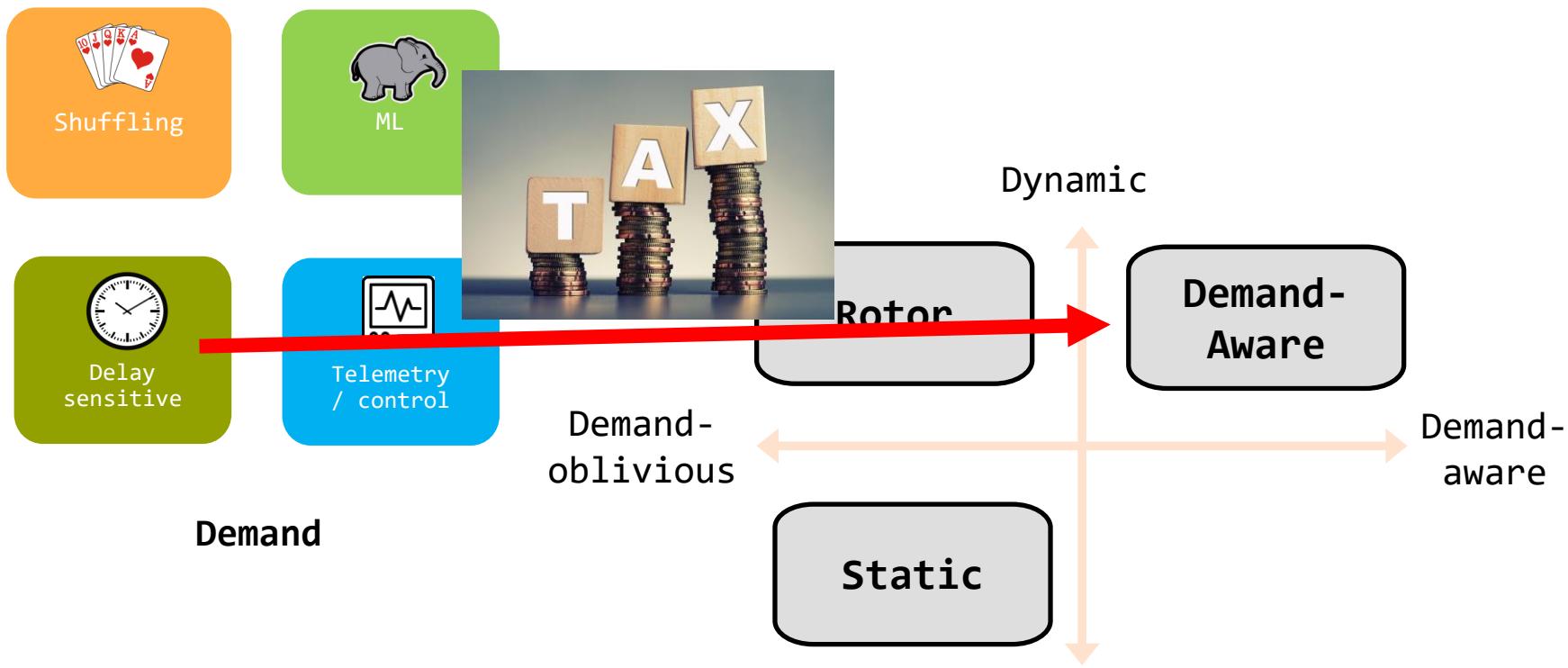
Examples:

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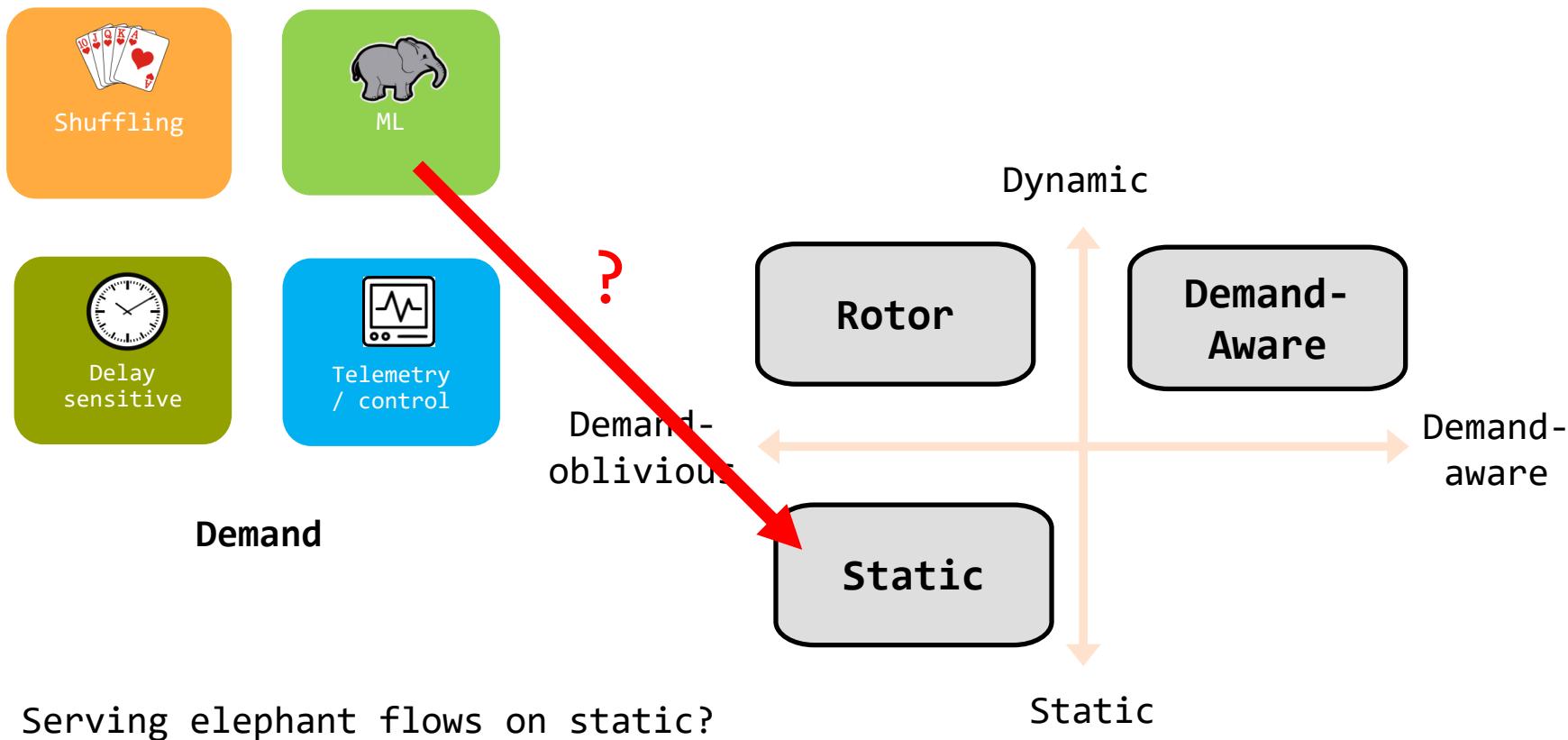
Match or Mismatch?



Serving mice flows on demand-aware?
Bad idea! Latency tax.

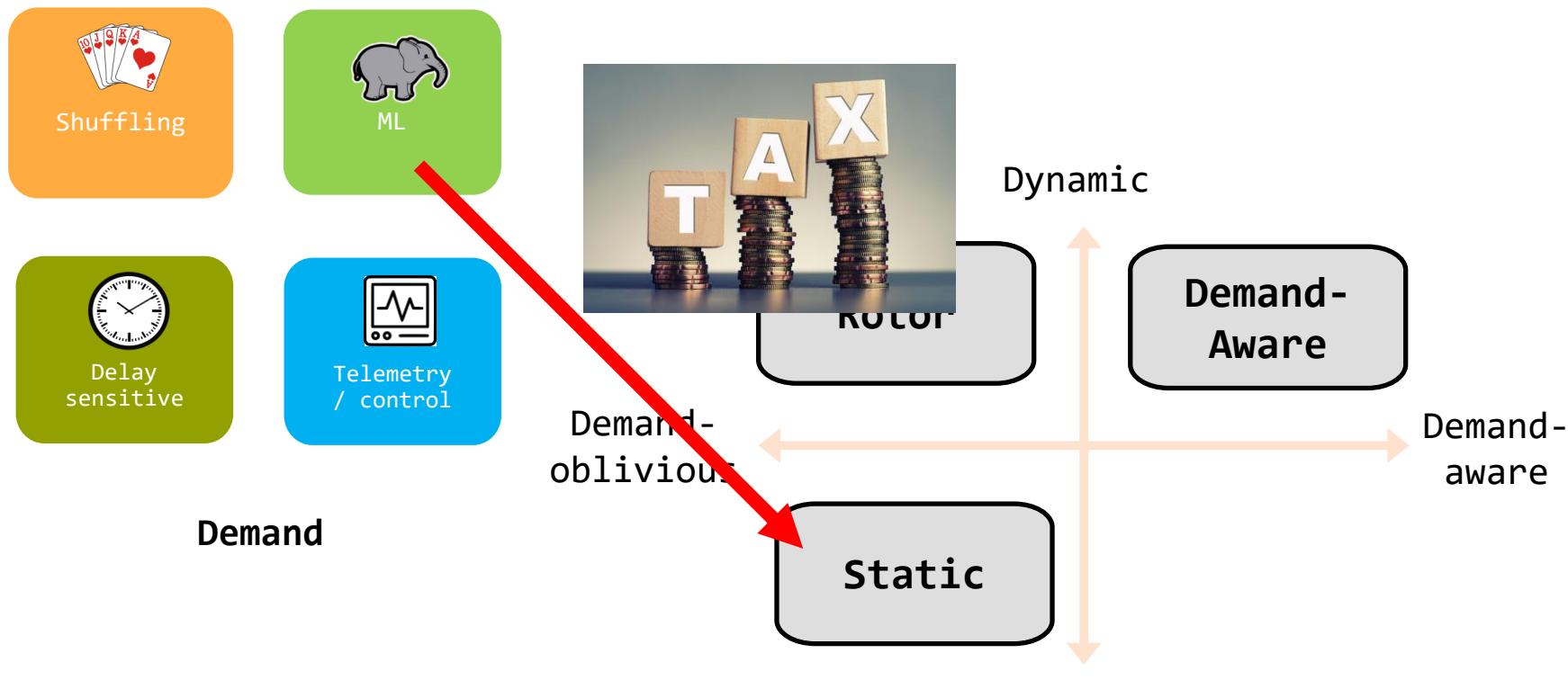
Examples:

Match or Mismatch?



Examples:

Match or Mismatch?



Serving elephant flows on static?
Bad idea! Bandwidth tax.

Examples: Match or Mismatch?



Demand

Demand-
oblivious

Dynamic

Demand-
aware

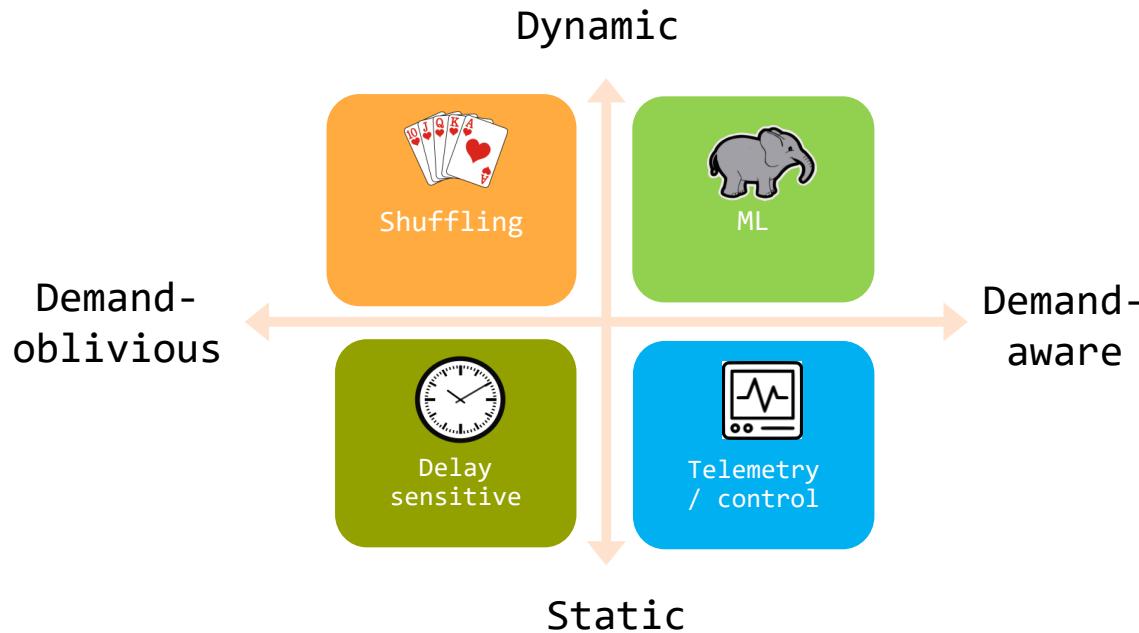
Static

Topology

Serving elephant flows on static?
Bad idea! Bandwidth tax.

Optimal Solution:

It's a  Match!



We have a first approach:

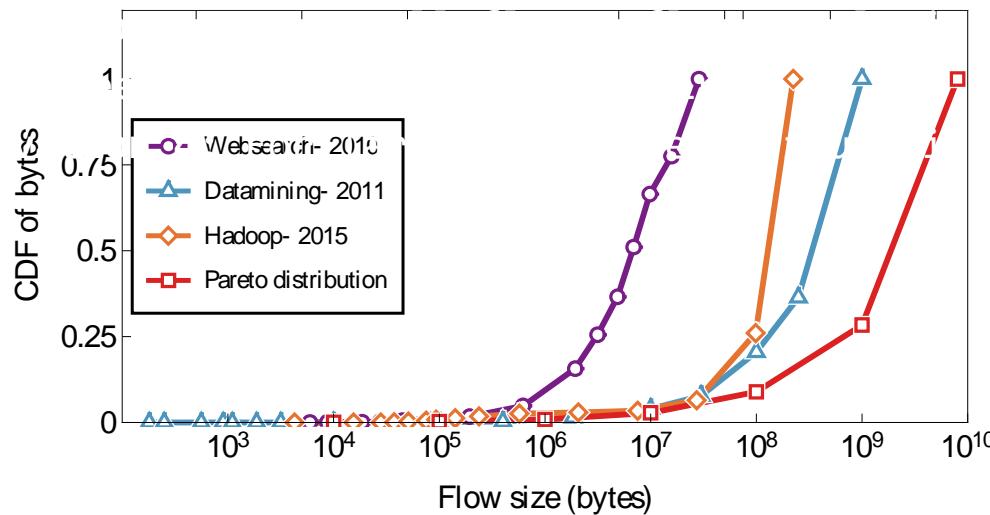
Cerberus* serves traffic on the “best topology”! (Optimality open)

Flow Size Matters

On what should topology type depend? We argue: **flow size.**

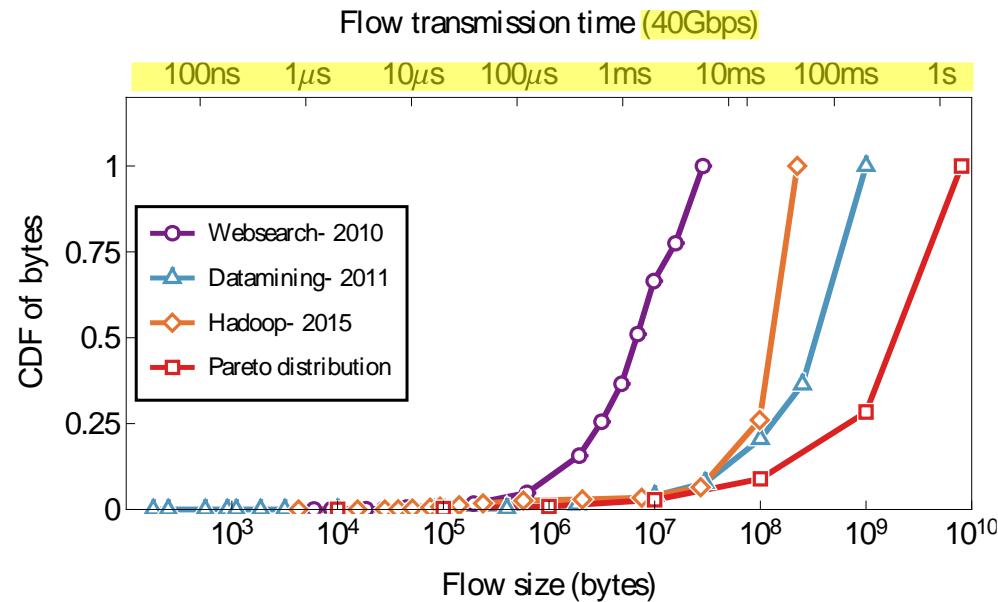
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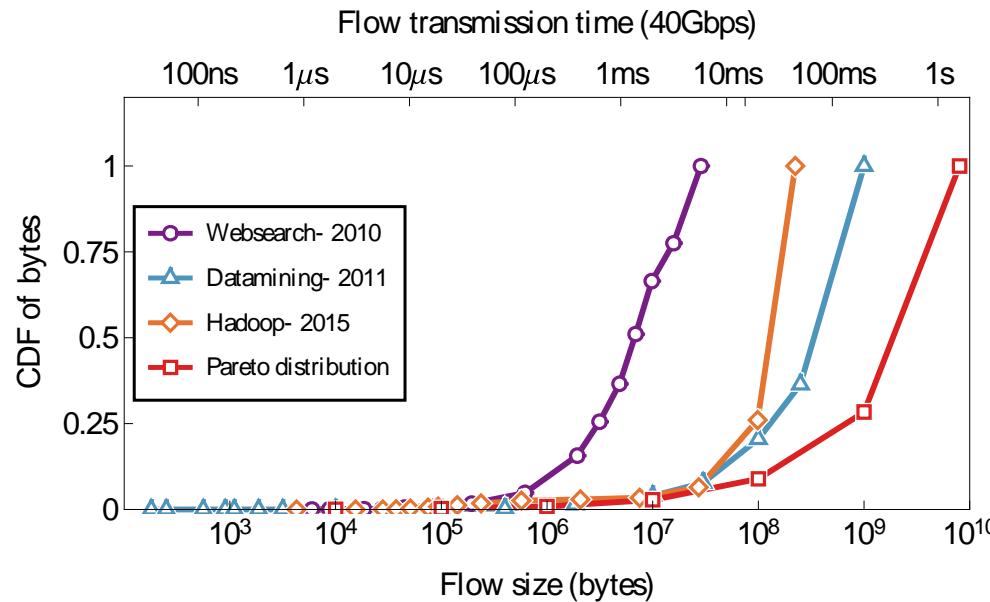
→ **Observation 1:** Different apps have different flow size distributions.

Flow Size Matters



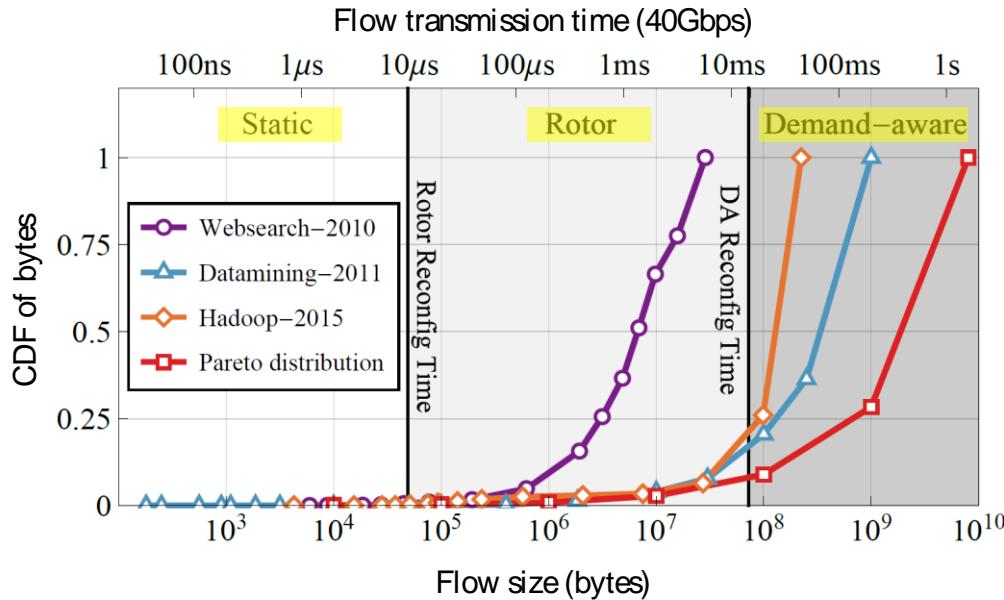
- **Observation 1:** Different apps have different flow size distributions.
- **Observation 2:** The transmission time of a flow depends on its **size**.

Flow Size Matters



- **Observation 1:** Different apps have different flow size distributions.
- **Observation 2:** The transmission time of a flow depends on its size.
- **Observation 3:** For small flows, flow completion time suffers if network needs to be reconfigured first.
- **Observation 4:** For large flows, reconfiguration time may amortize.

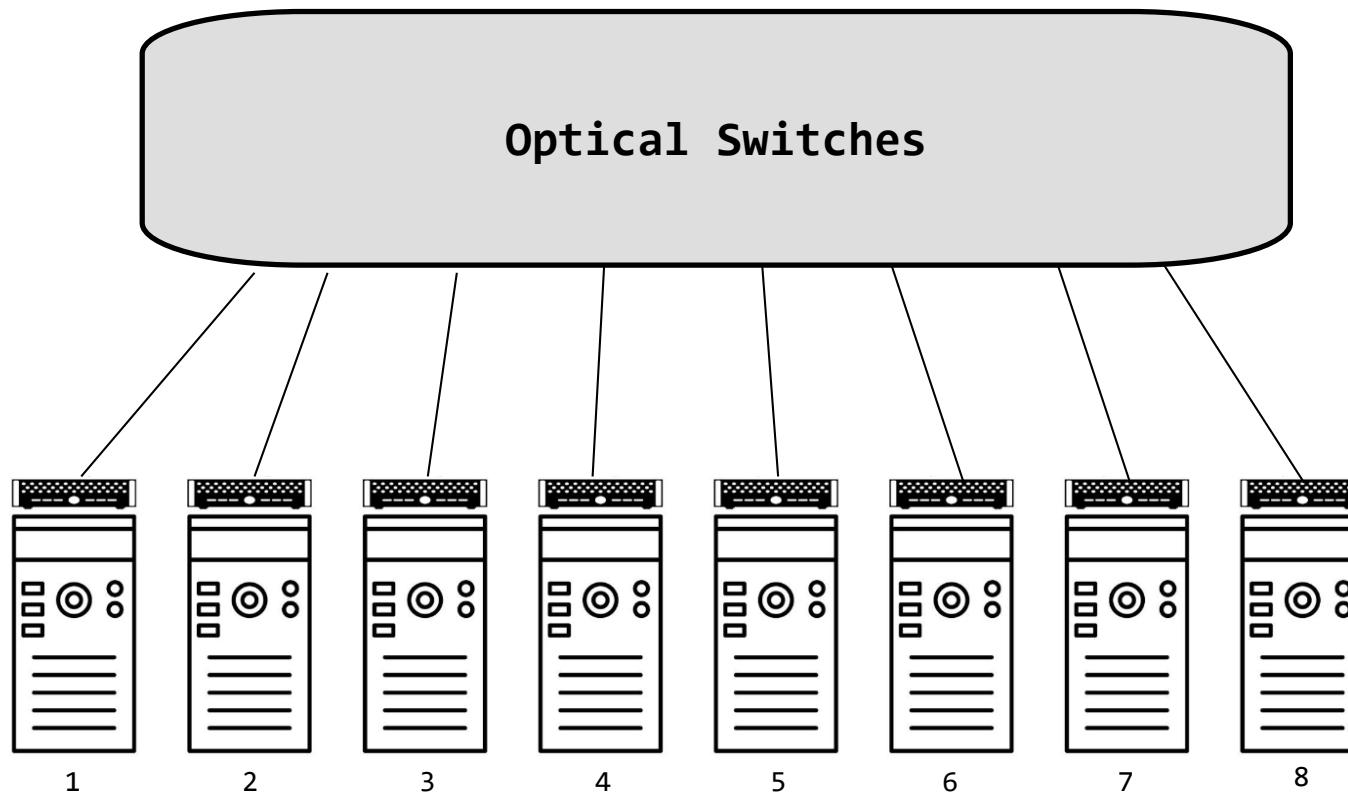
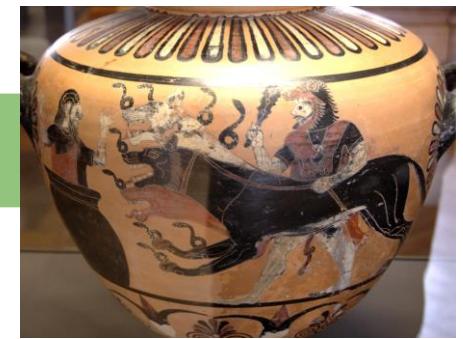
Flow Size Matters



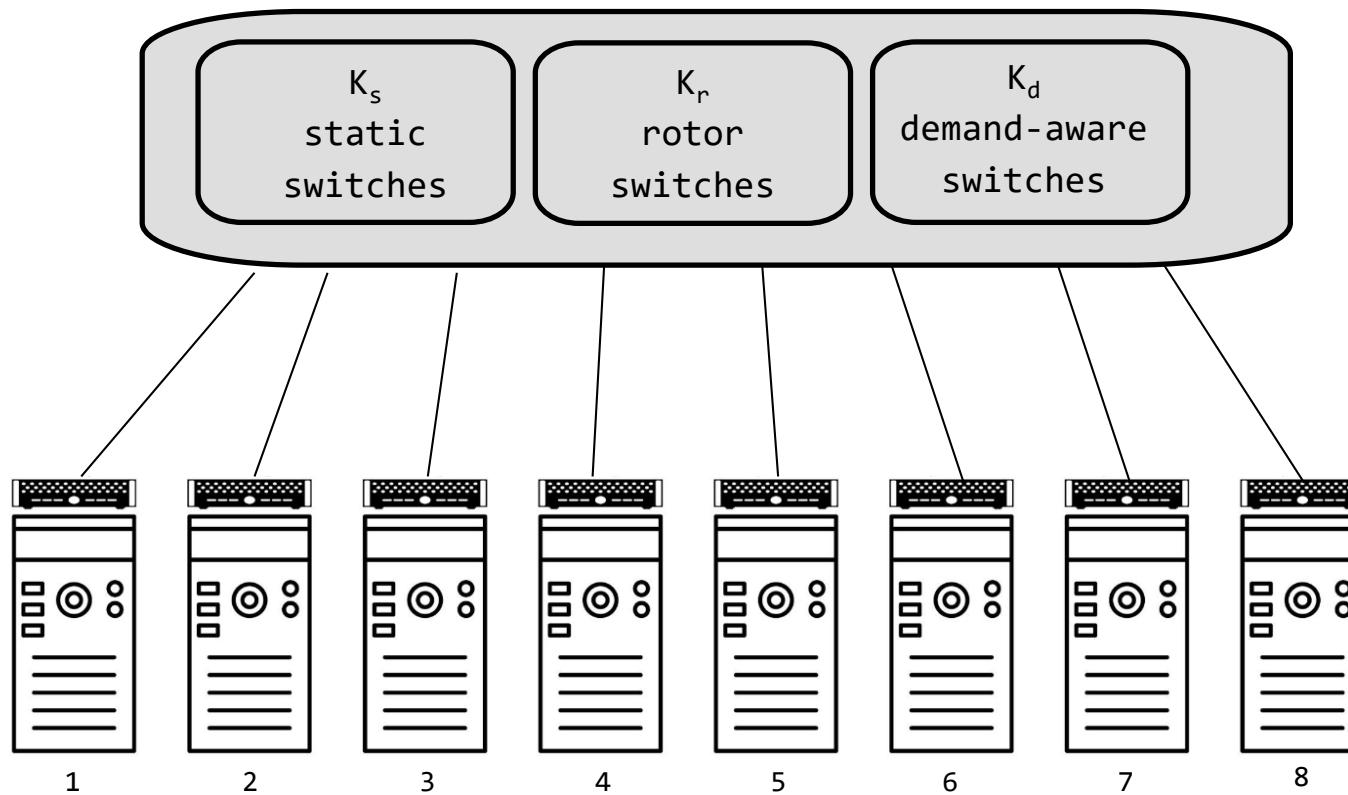
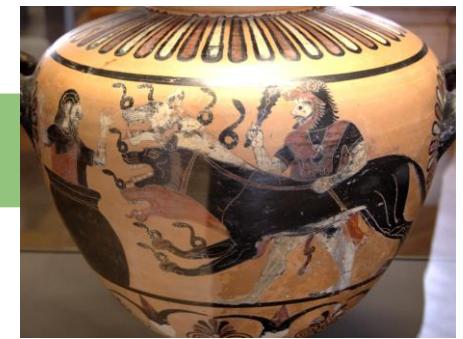
It's a 🔥 Match!

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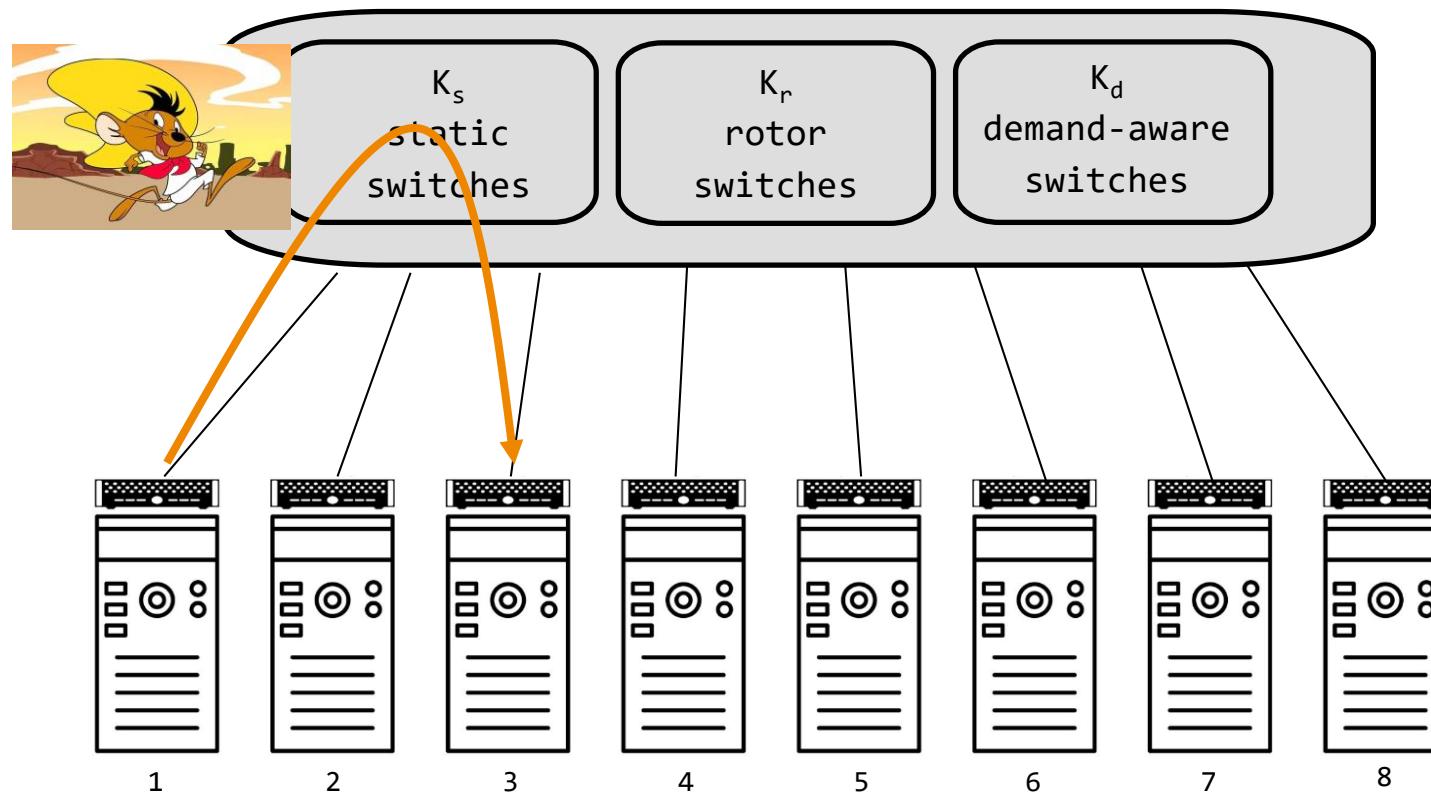
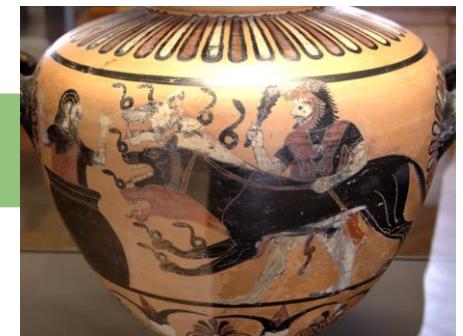
Cerberus



Cerberus

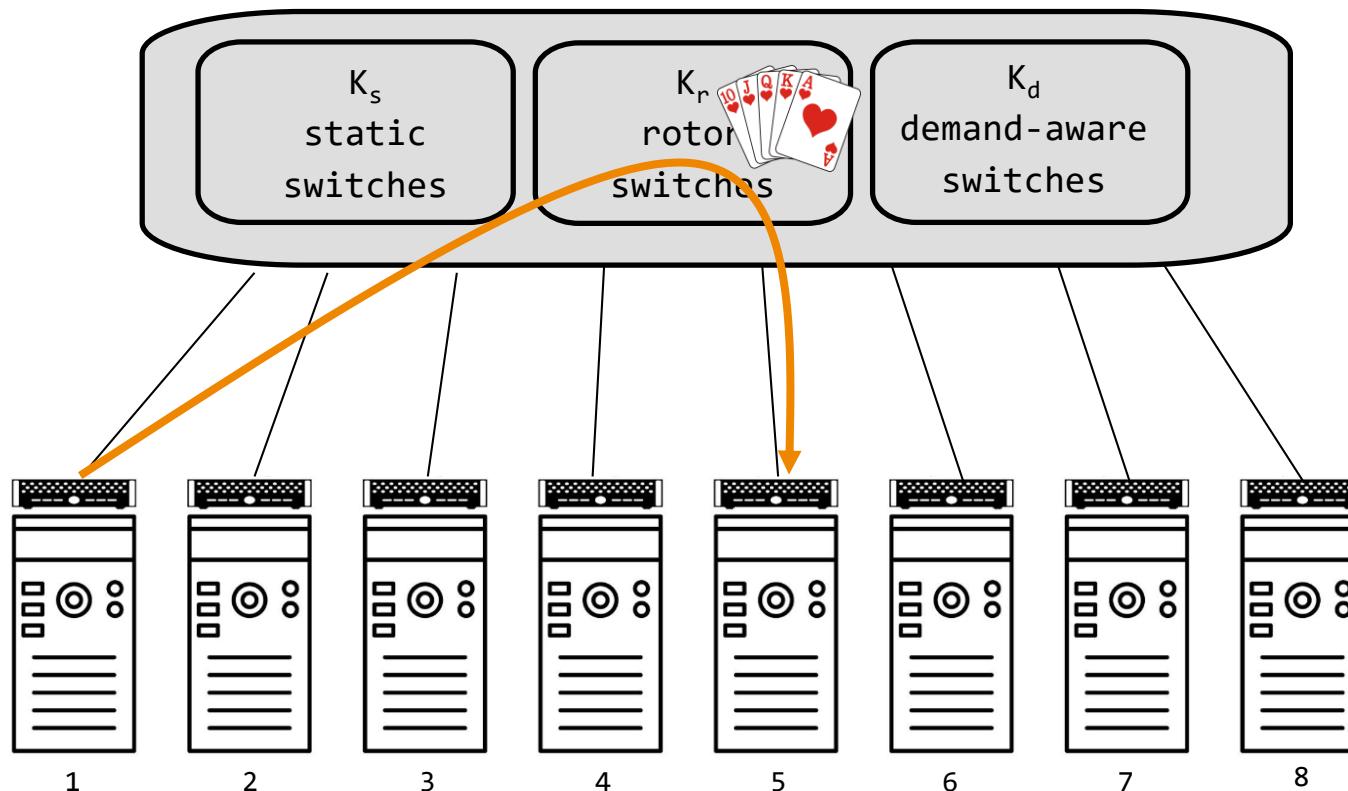
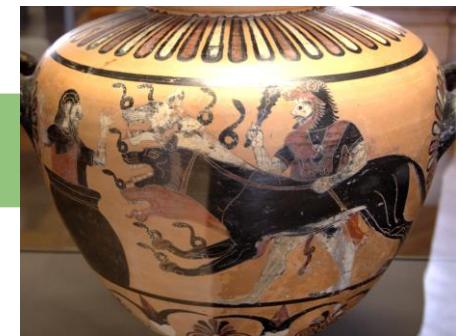


Cerberus



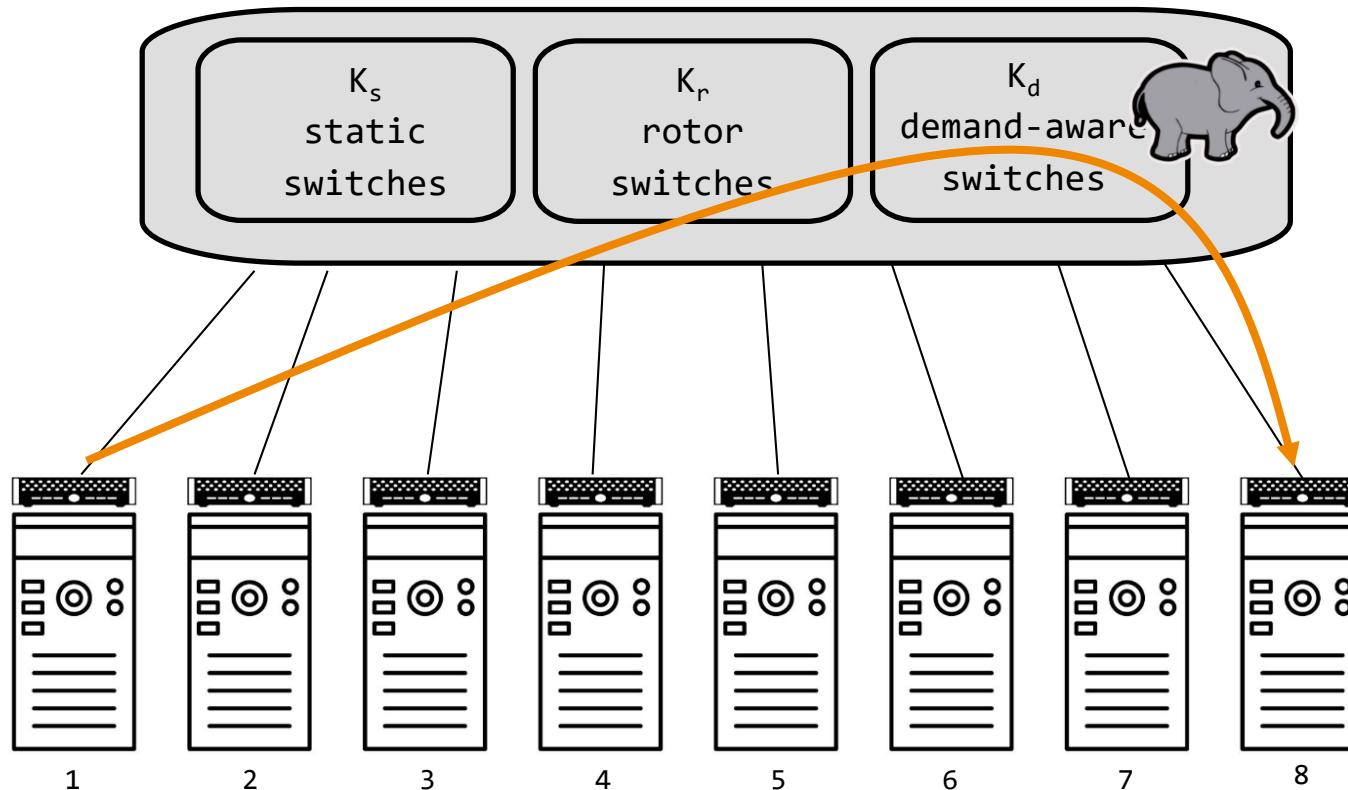
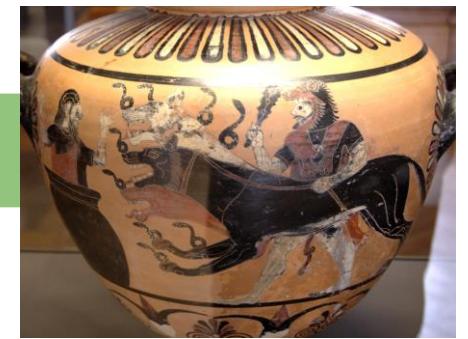
Scheduling: Small flows go via static switches...

Cerberus



Scheduling: ... **medium flows** via rotor switches...

Cerberus



Scheduling: ... and **large flows** via demand-aware switches
(if one available, otherwise via rotor).

Roadmap

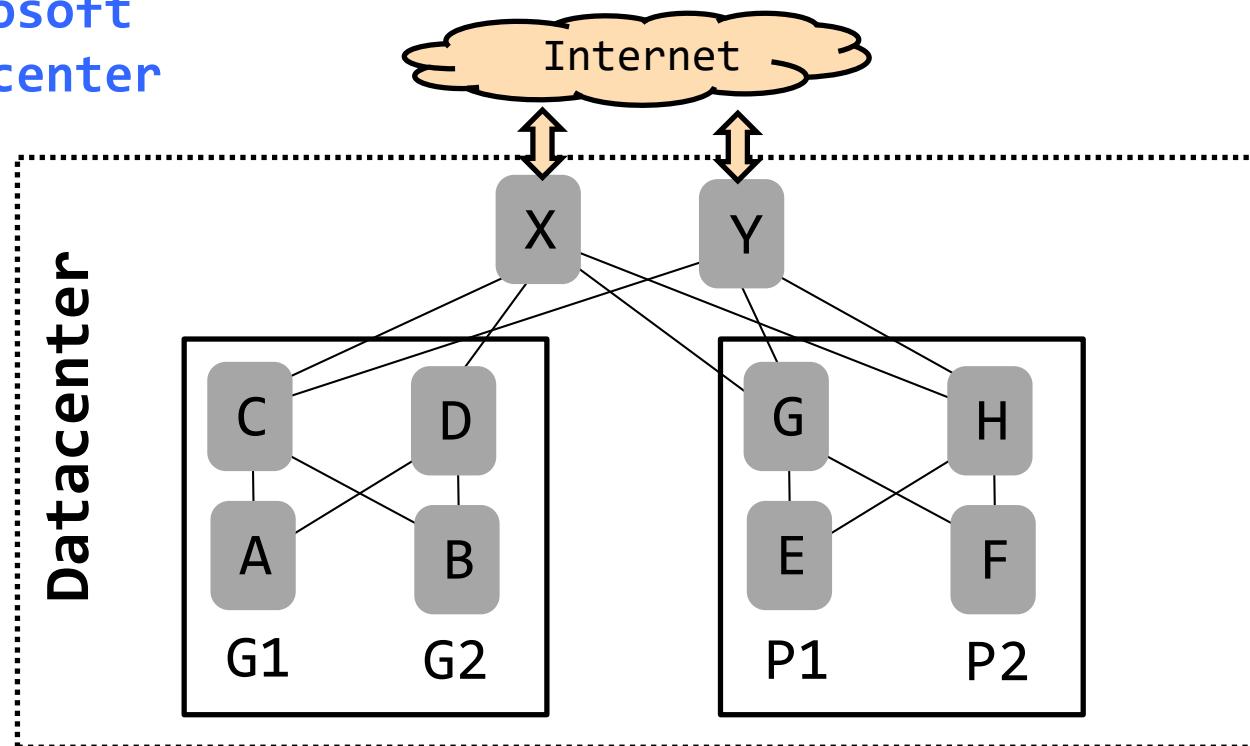


- Performance: Self-adjusting datacenter networks
- Modelling: How to model workloads, such as ML workloads?
- Dependability: Self-correcting MPLS networks
- More Use cases for self-driving networks

Challenge: Complexity

Especially Under Failures (Policy Compliance)

Example: BGP in
Microsoft
datacenter

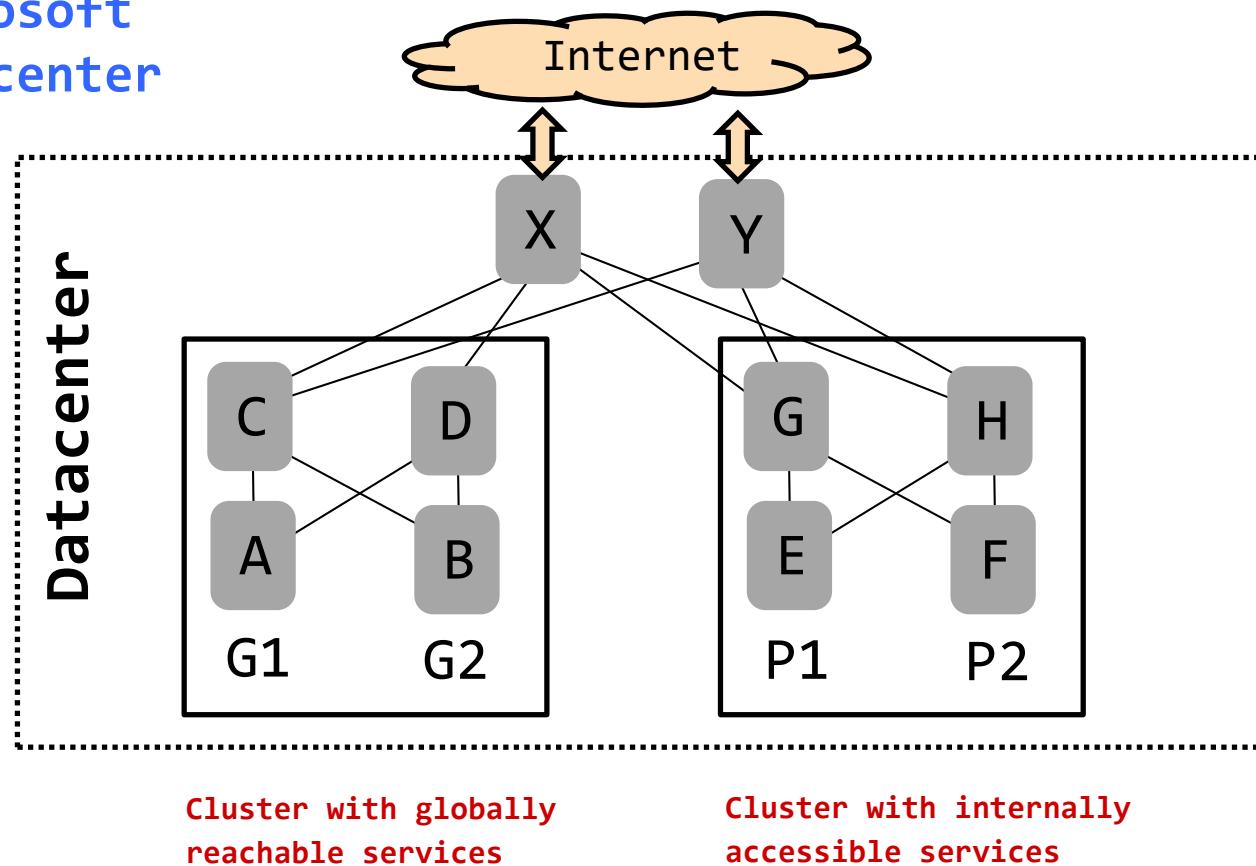


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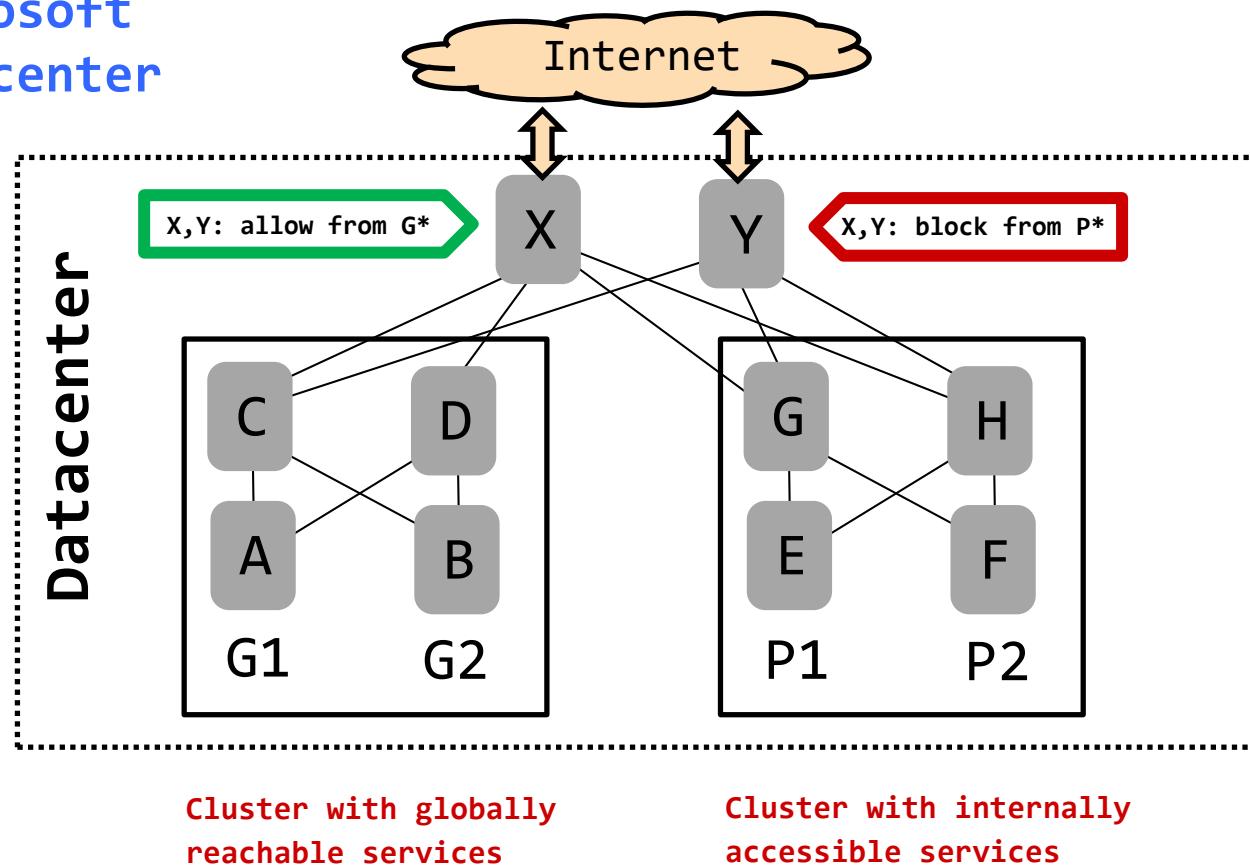
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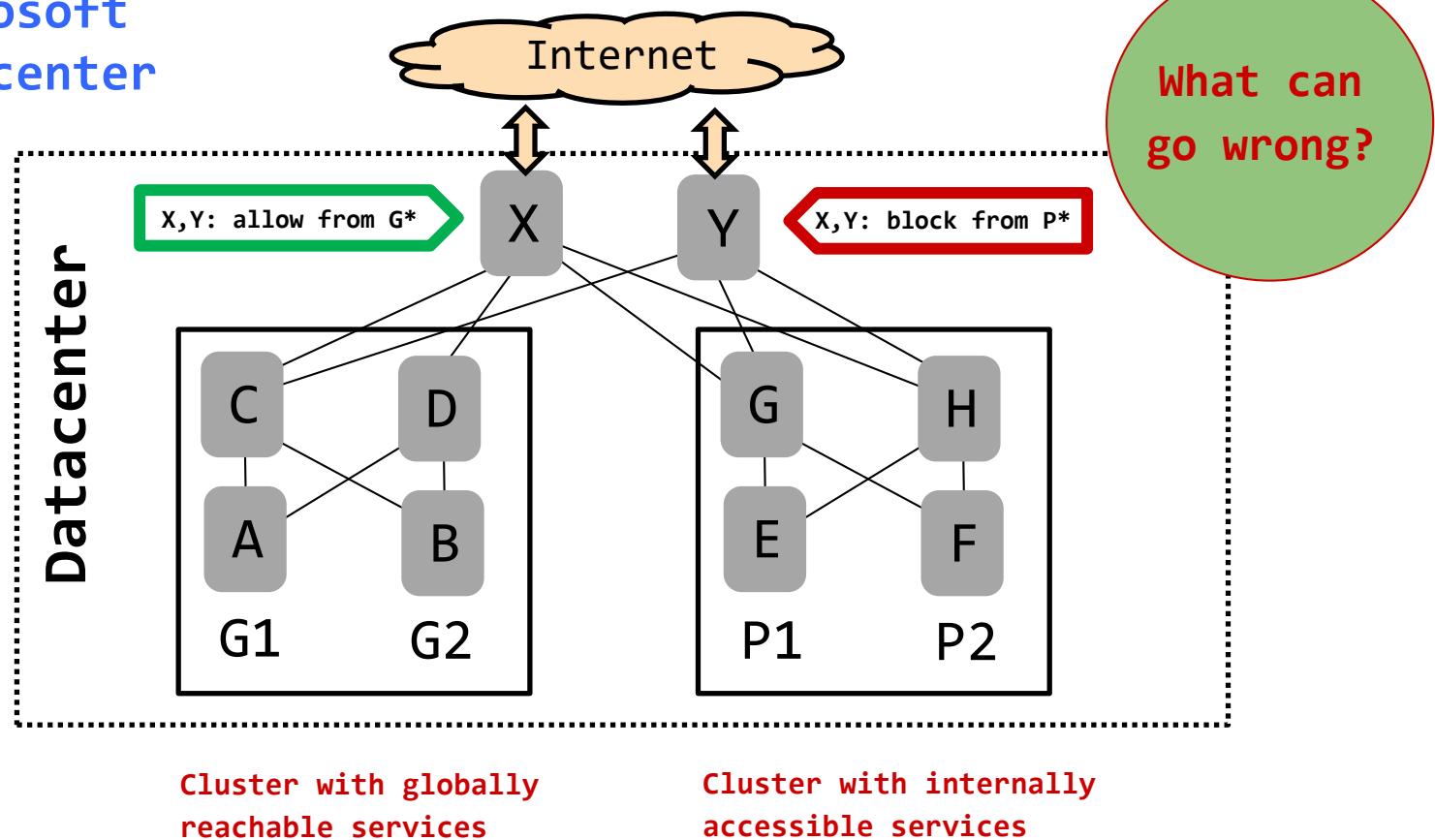
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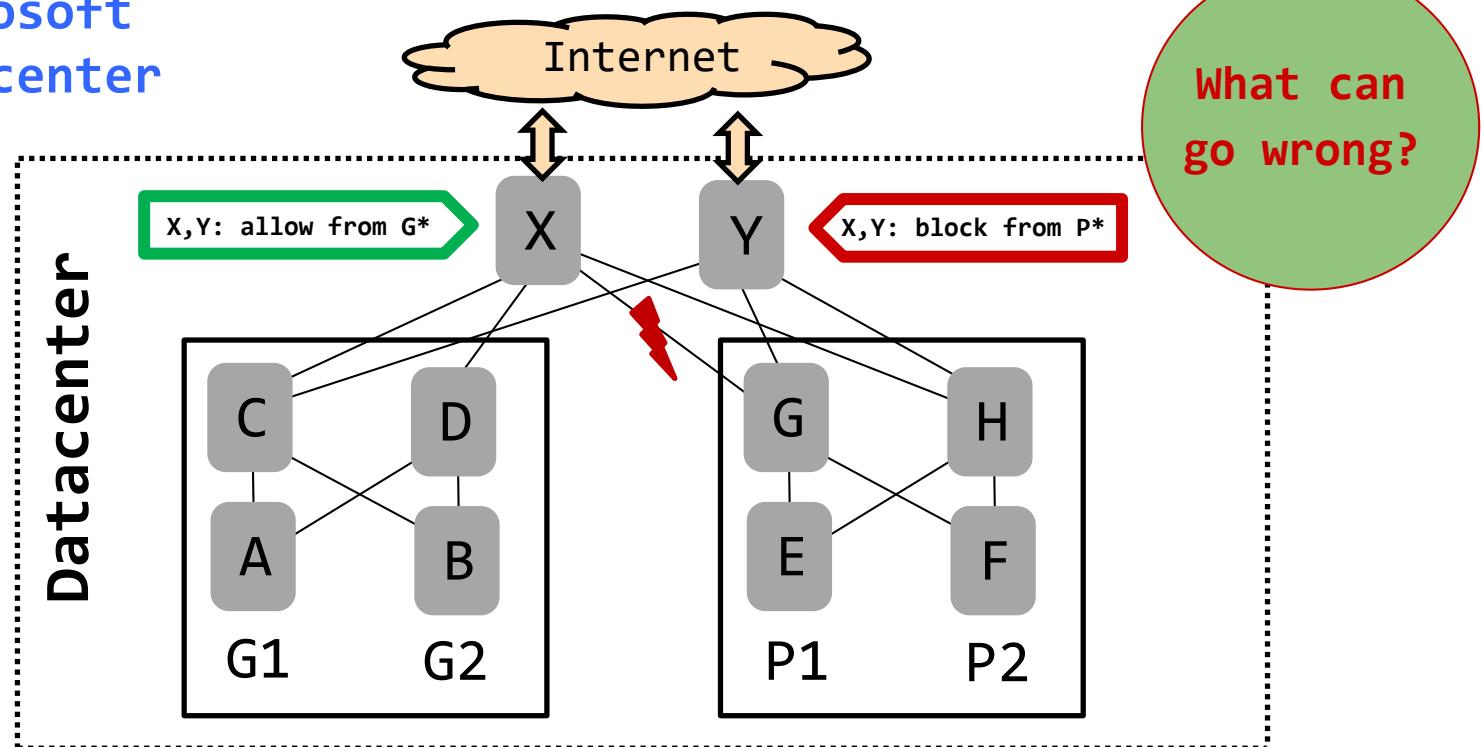
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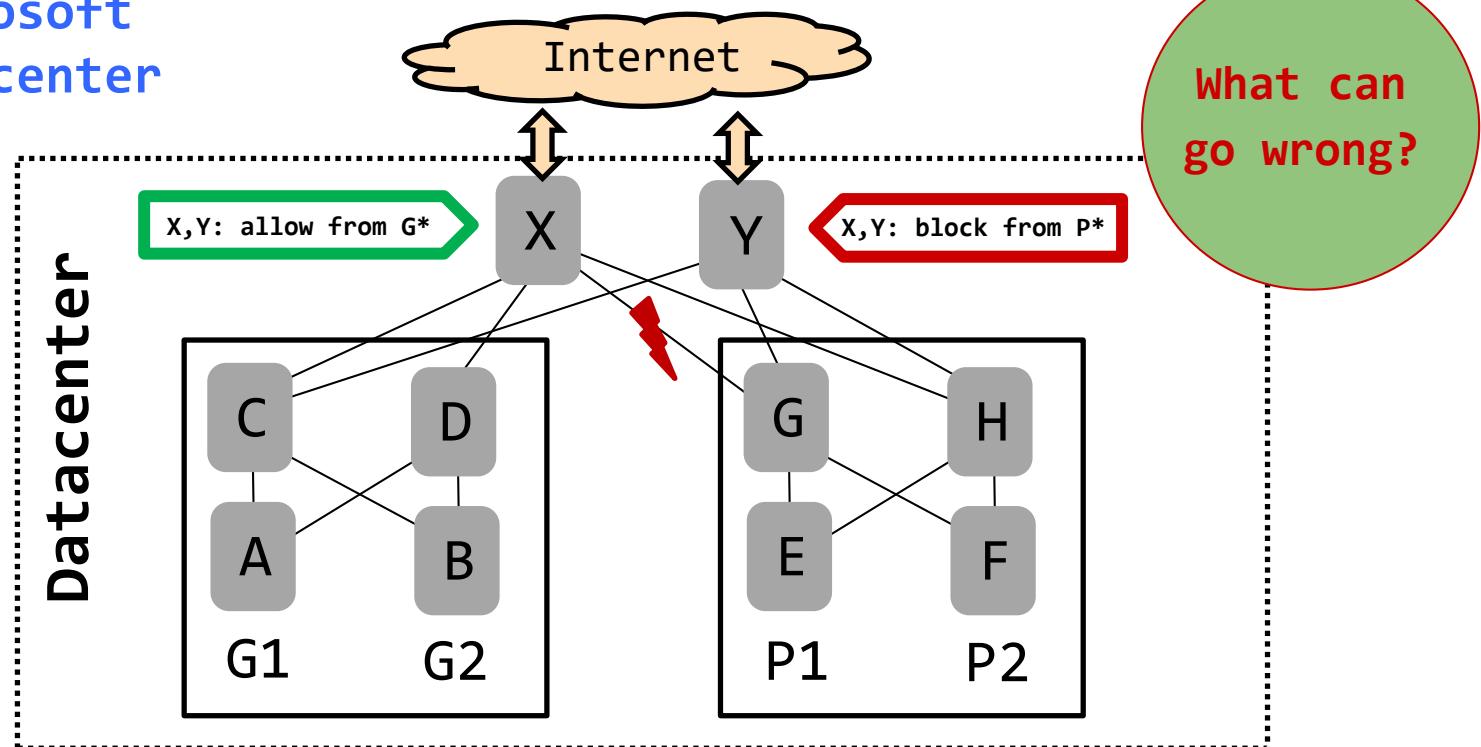
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Microsoft
datacenter



Challenge: Complexity

Especially Under Failures (Policy Compliance)

Example: BGP in
Microsoft
datacenter

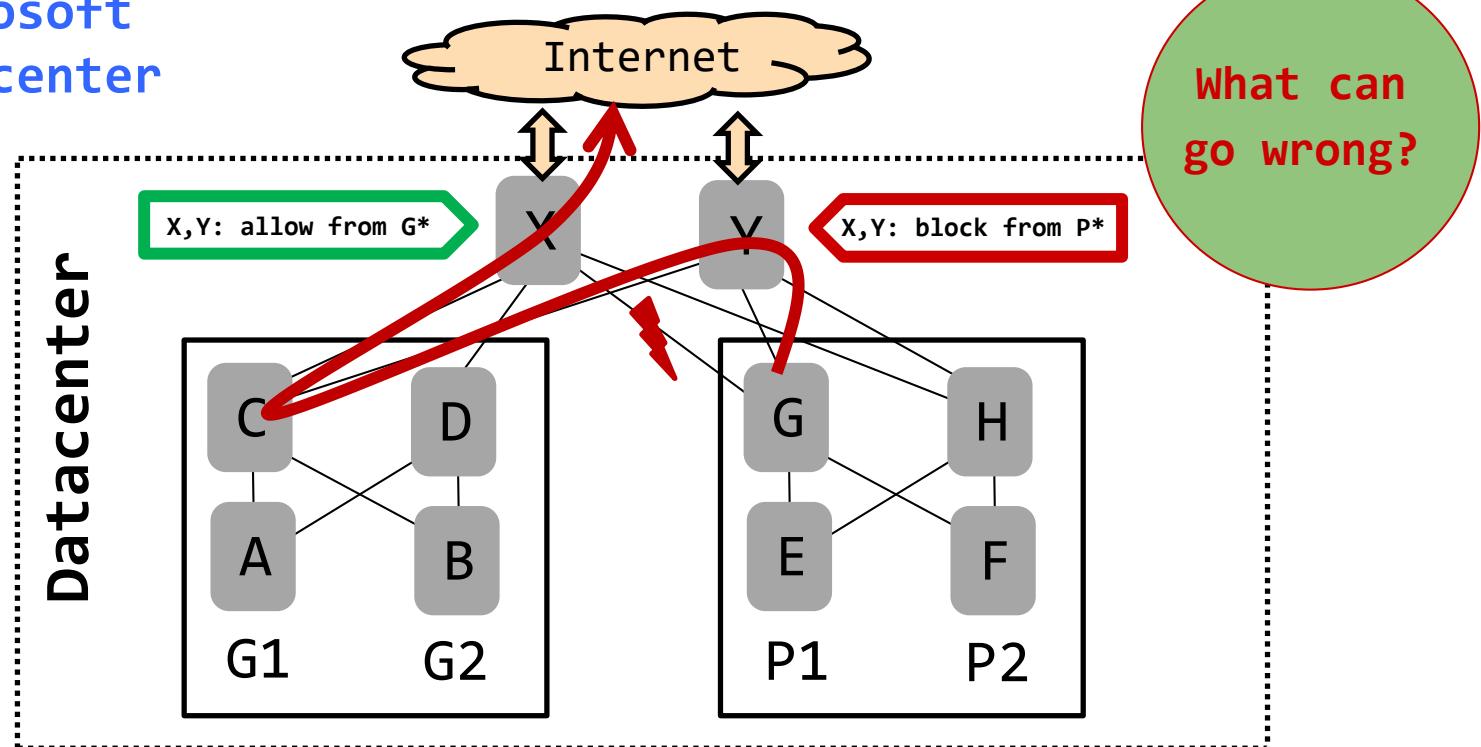


If link (G,X) fails and traffic from G is rerouted via Y and C to X:
X announces (does not block) G and H as it comes from C. (Note: BGP.)

Challenge: Complexity

Especially Under Failures (Policy Compliance)

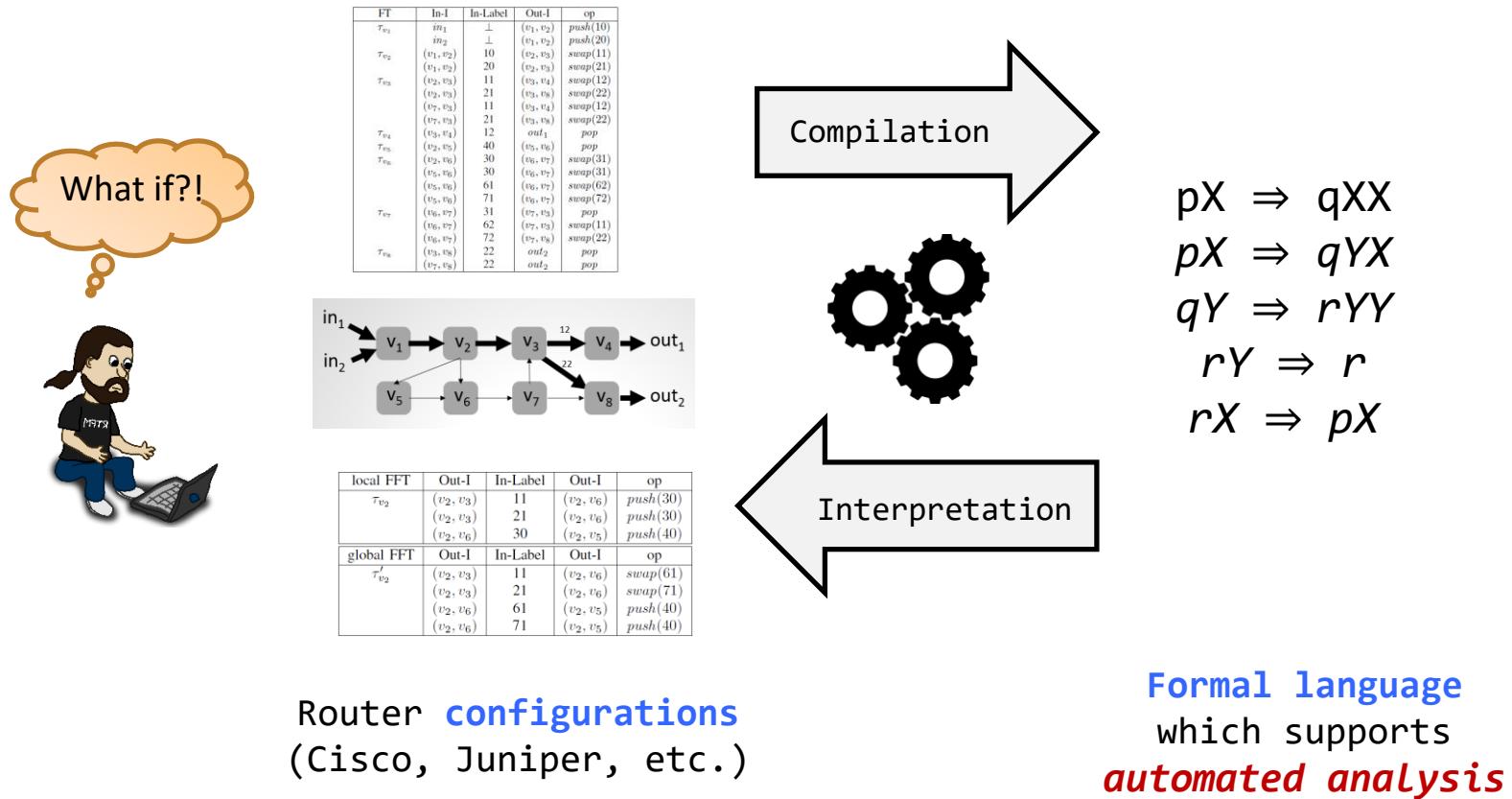
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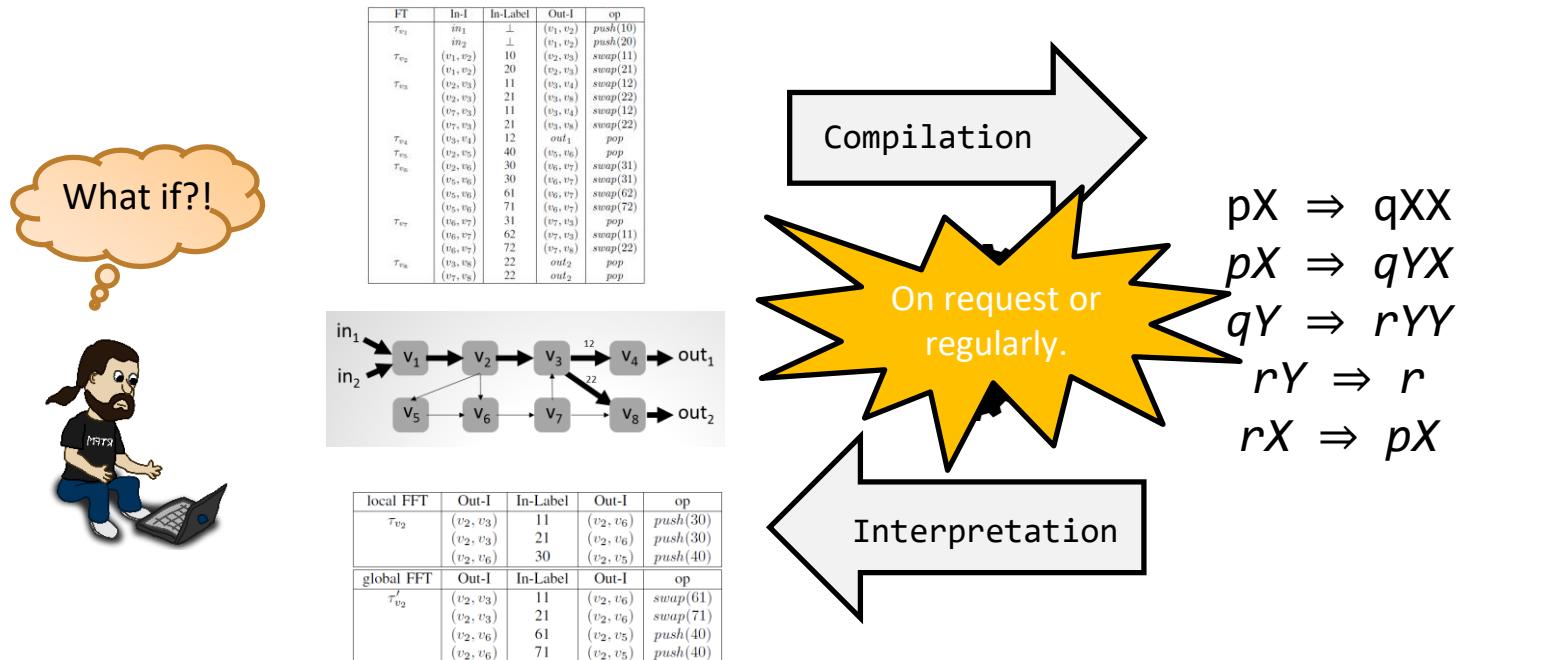
Dependable Networks with Automated Whatif Analysis

- Formal methods good for verifying networks! E.g., P-Rex for MPLS (Jensen et al. CoNEXT'19)



Dependable Networks with Automated Whatif Analysis

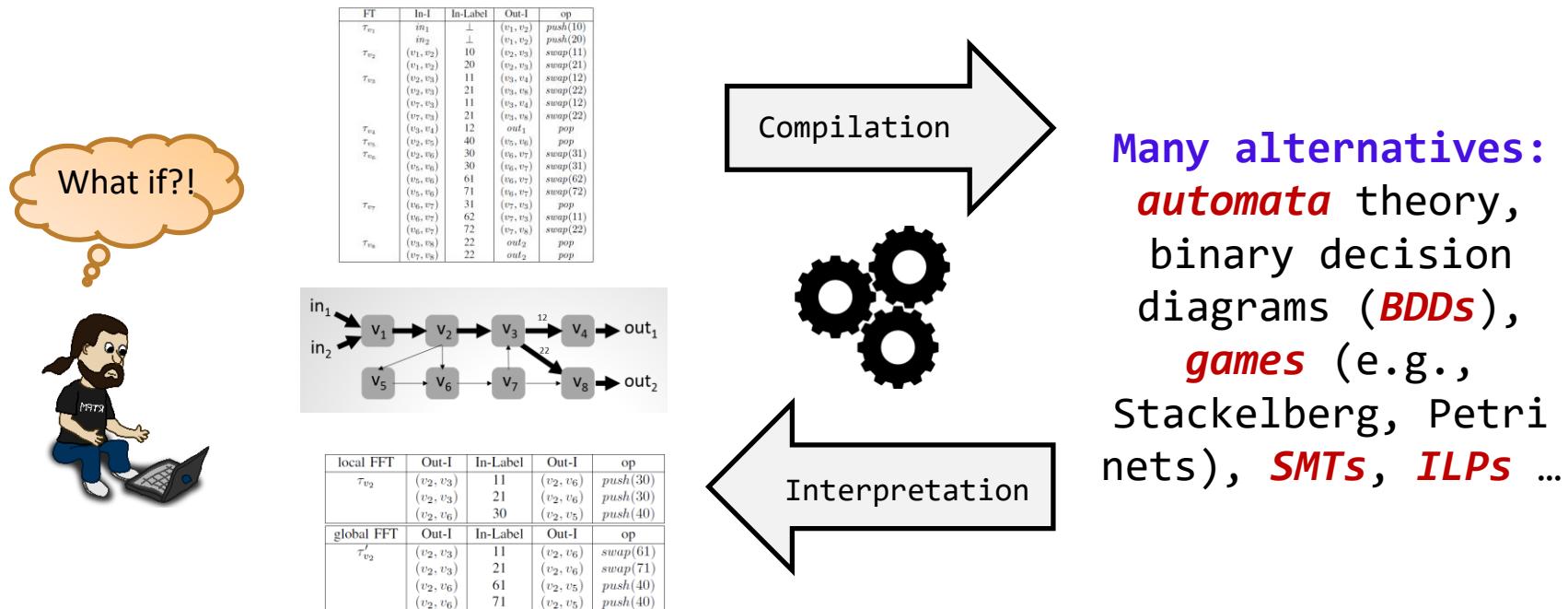
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Formal language
which supports
automated analysis

Dependable Networks with Automated Whatif Analysis

- Formal methods good for verifying networks! E.g., P-Rex for MPLS (Jensen et al. CoNEXT'19)



Even more automation:

Synthesis

- Formal methods good for verifying networks! E.g., P-Rex for MPLS (Jensen et al. CoNEXT'19)

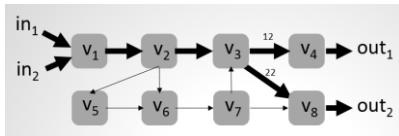


FT	In-I	In-Label	Out-I	op
τ_{v_1}	in_1		(v_1, v_2)	$push(10)$
	in_2	\perp	(v_1, v_2)	$push(20)$
τ_{v_2}	(v_1, v_2)	10	(v_2, v_3)	$swap(11)$
	(v_1, v_3)	20	(v_2, v_3)	$swap(21)$
τ_{v_3}	(v_2, v_3)	11	(v_3, v_4)	$swap(12)$
	(v_2, v_3)	21	(v_3, v_4)	$swap(22)$
τ_{v_4}	(v_3, v_4)	11	(v_3, v_4)	$swap(12)$
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τ_{v_5}	(v_3, v_4)	12	out_1	pop
	(v_3, v_4)	40	(v_3, v_4)	pop
τ_{v_6}	(v_2, v_6)	30	(v_4, v_7)	$swap(31)$
	(v_5, v_6)	30	(v_4, v_7)	$swap(31)$
τ_{v_7}	(v_5, v_6)	61	(v_4, v_7)	$swap(62)$
	(v_5, v_6)	71	(v_4, v_7)	$swap(72)$
τ_{v_8}	(v_6, v_7)	31	(v_7, v_8)	pop
	(v_6, v_7)	62	(v_7, v_8)	$swap(11)$
τ_{v_9}	(v_6, v_7)	72	(v_7, v_8)	$swap(22)$
	(v_3, v_8)	22	out_2	pop
τ_{v_9}	(v_7, v_8)	22	out_2	pop

Compilation



*Where configuration
not compliant?*



local FFT	Out-I	In-Label	Out-I	op
τ_{v_2}	(v_2, v_3)	11	(v_2, v_6)	$push(30)$
	(v_2, v_3)	21	(v_2, v_6)	$push(30)$
	(v_2, v_3)	30	(v_2, v_5)	$push(40)$
global FFT	Out-I	In-Label	Out-I	op
τ'_{v_2}	(v_2, v_3)	11	(v_2, v_6)	$swap(61)$
	(v_2, v_3)	21	(v_2, v_6)	$swap(71)$
	(v_2, v_6)	61	(v_2, v_5)	$push(40)$
	(v_2, v_6)	71	(v_2, v_5)	$push(40)$

Synthesis!



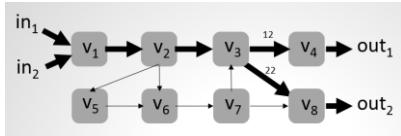
Router configurations
(Cisco, Juniper, etc.)

Even more automation: Synthesis

- Formal methods good for verifying networks! E.g., P-Rex for MPLS (Jensen et al. CoNEXT'19)

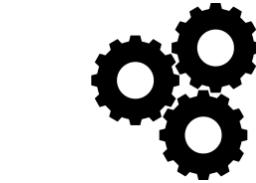
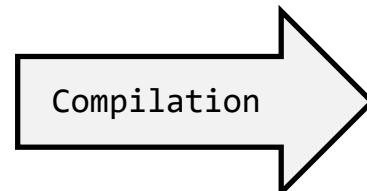


FT	In-I	In-Label	Out-I	op
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Router configurations
(Cisco, Juniper, etc.)



*Where configuration
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Even more automation:

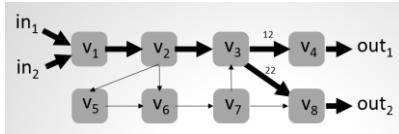
Synthesis

- Formal methods good for verifying networks! E.g., P-Rex for MPLS (Jensen et al. CoNEXT'19)



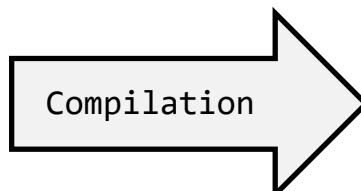
All will be fine!

FT	In-I	In-Label	Out-I	op
τ_{v_1}	in_1		(v_1, v_2)	$push(10)$
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Router configurations
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Compilation



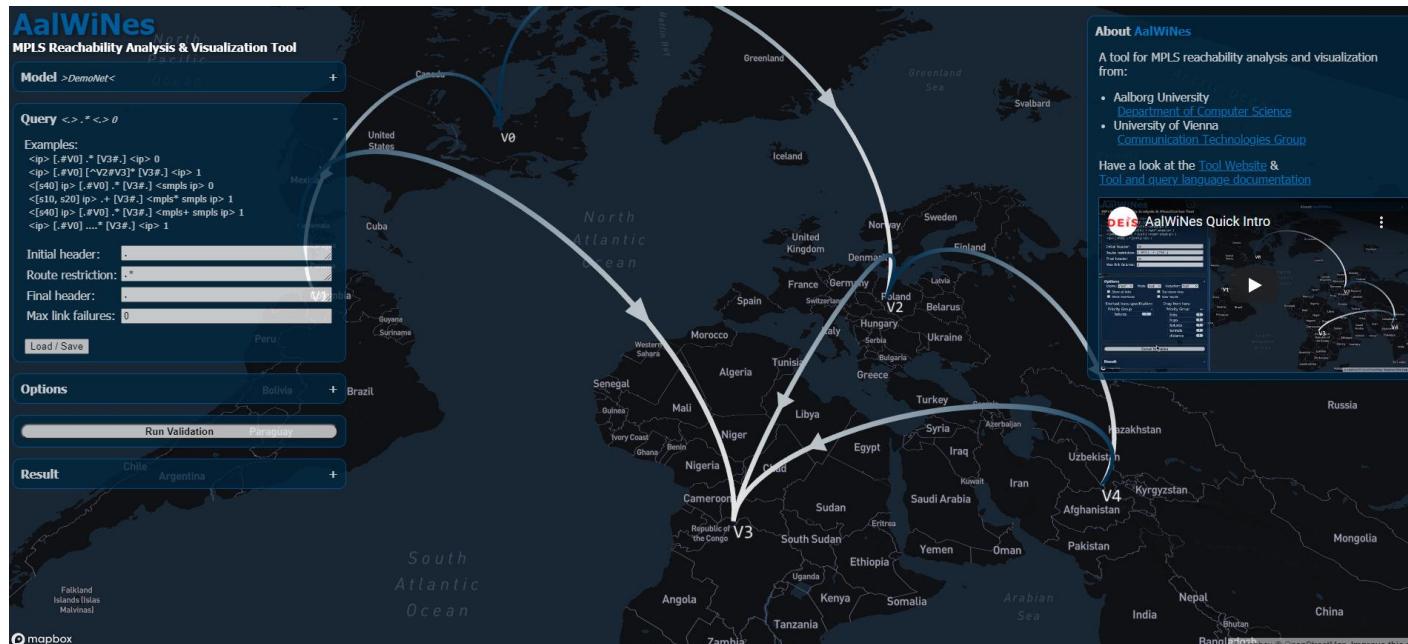
Where configuration
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Synthesis!



P-Rex / AalWiNes Tool



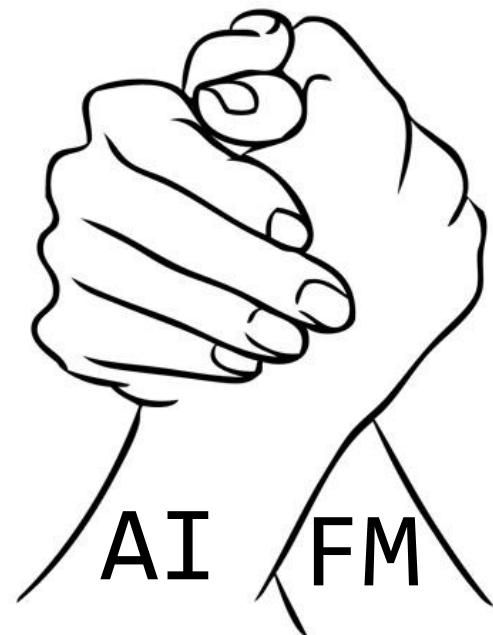
Tool: <https://demo.aalwines.cs.aau.dk/>

Youtube: https://www.youtube.com/watch?v=mvXAn9i7_00

Efficient Synthesis? ML+FM!



- Formal *synthesis slower* than verification
- An opportunity for using ML!
- *Ideally ML+FM*: guarantees from formal methods, performance from ML
- For example: synthesize with ML then verify with formal methods
- Examples: DeepMPLS, DeepBGP, ...



On a related note:

Human in the Loop?

- When and how to keep the human in the loop?
- Critical: can system realize when *help* is needed?
- But AI tools (e.g. LLM) may also influence the human: can the operator become *too confident* with such tools?
- Challenges to be discussed!

Roadmap



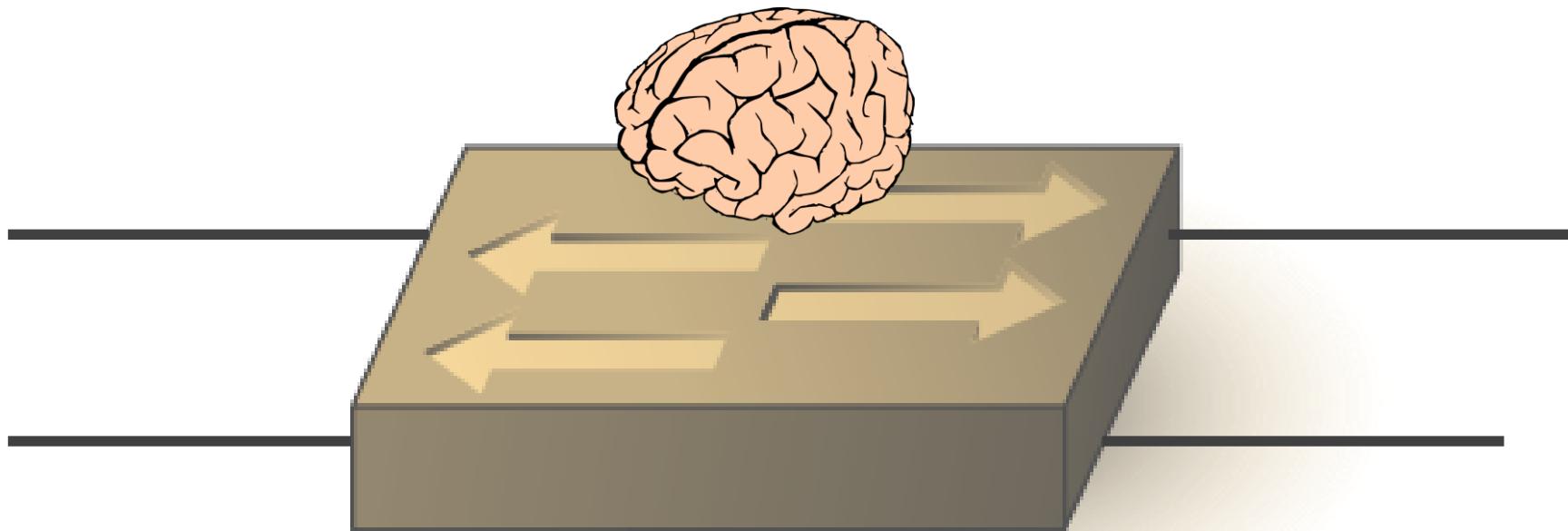
- Performance: Self-adjusting datacenter networks
- Modelling: How to model workloads, such as ML workloads?
- Dependability: Self-correcting MPLS networks
- More use cases for self-driving networks

Smart Switches



Smart Switches

→ What if switches become smart?



Scenario 1

→ What if switches become smart? Assume: shared memory size 3.



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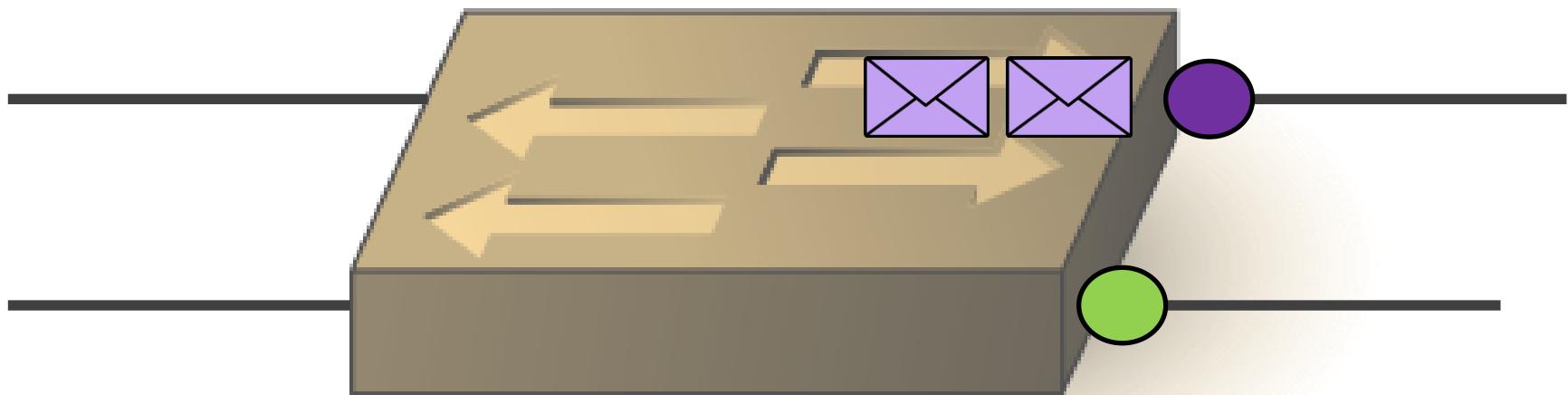
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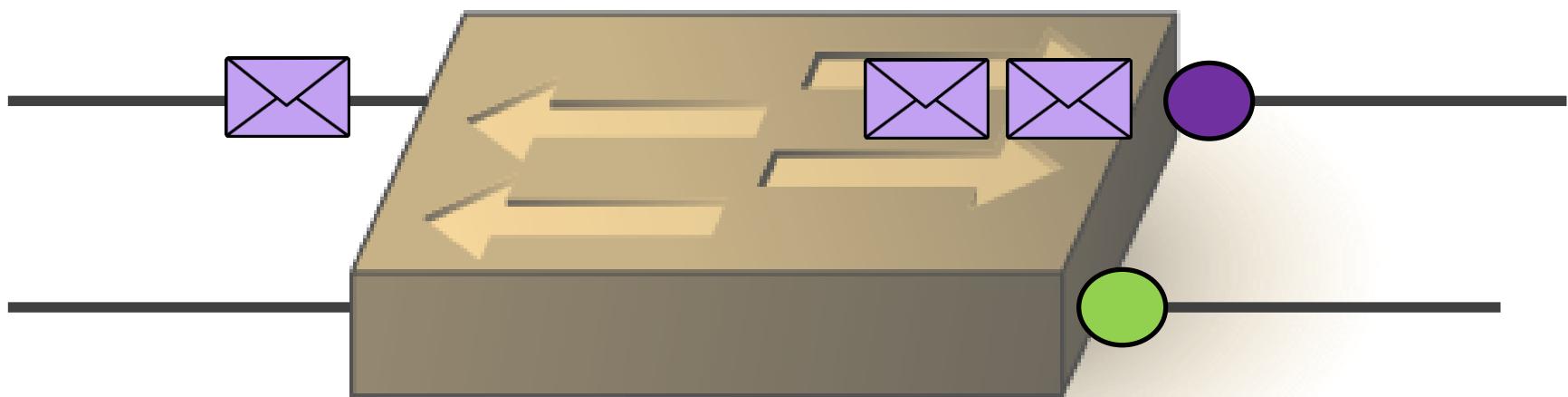
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Scenario 1

→ What if switches become smart? Assume: shared memory size 3.



Scenario 1

- What if switches become smart? Assume: shared memory size 3.



- Suboptimal: green packets could be transmitted in parallel, but there is no more space! (Output rate 1 vs 2!)

Scenario 2

→ What if switches become smart? Assume: shared memory size 3.



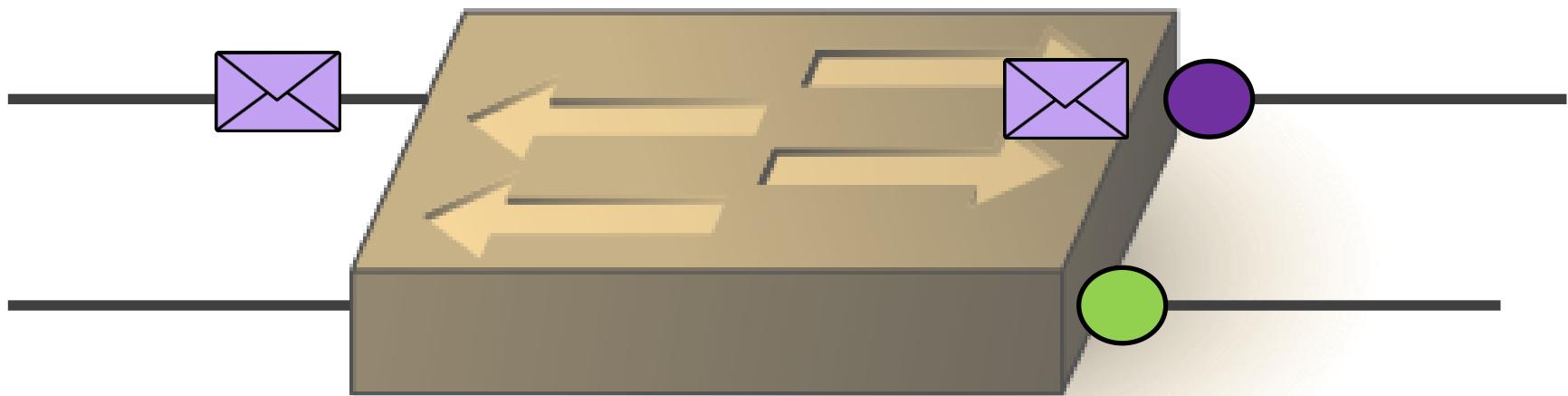
Scenario 2

→ What if switches become smart? Assume: shared memory size 3.



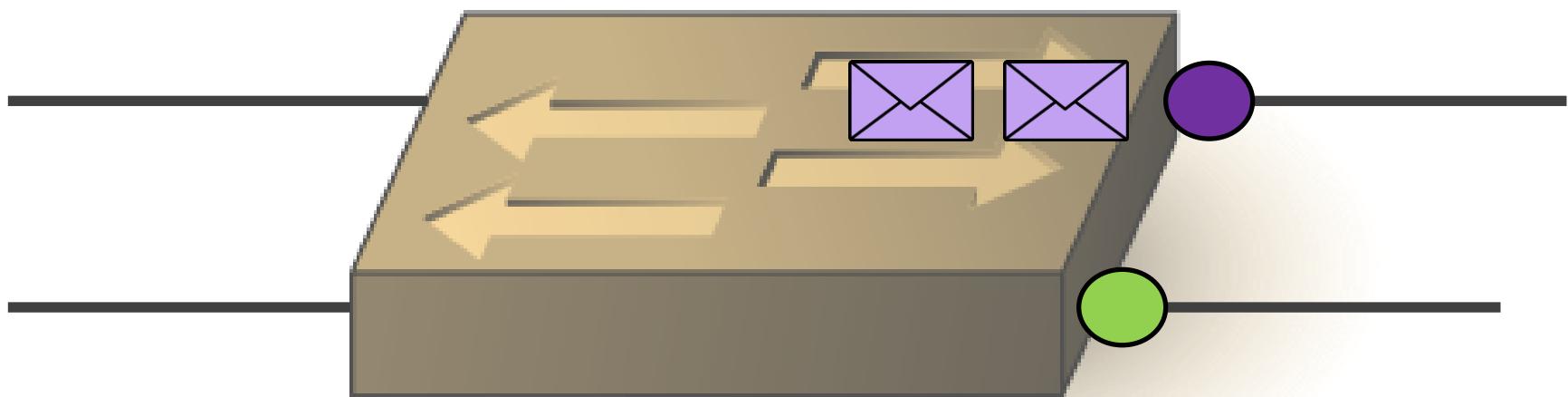
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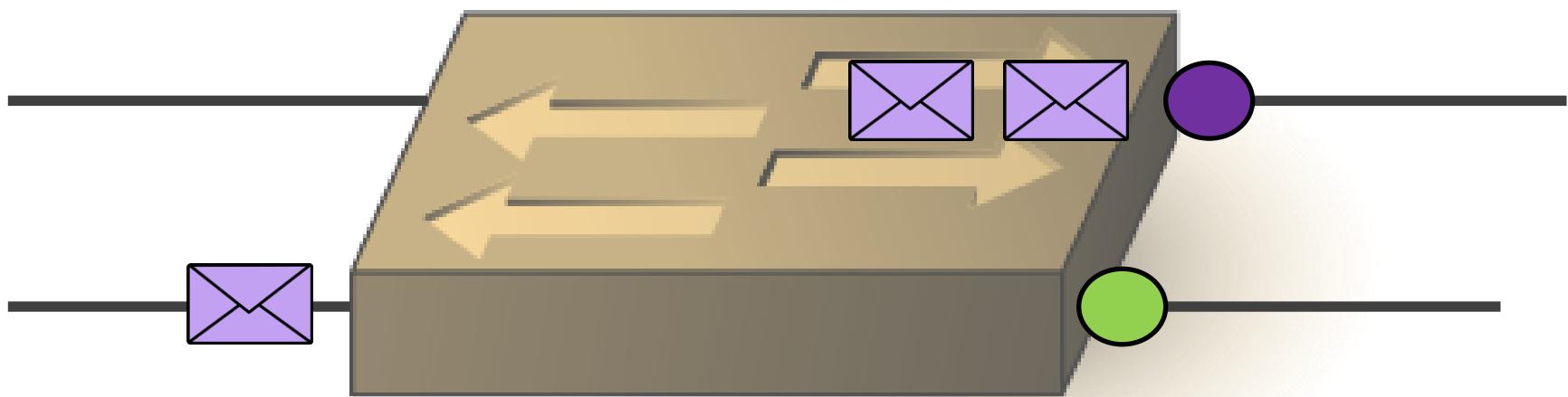
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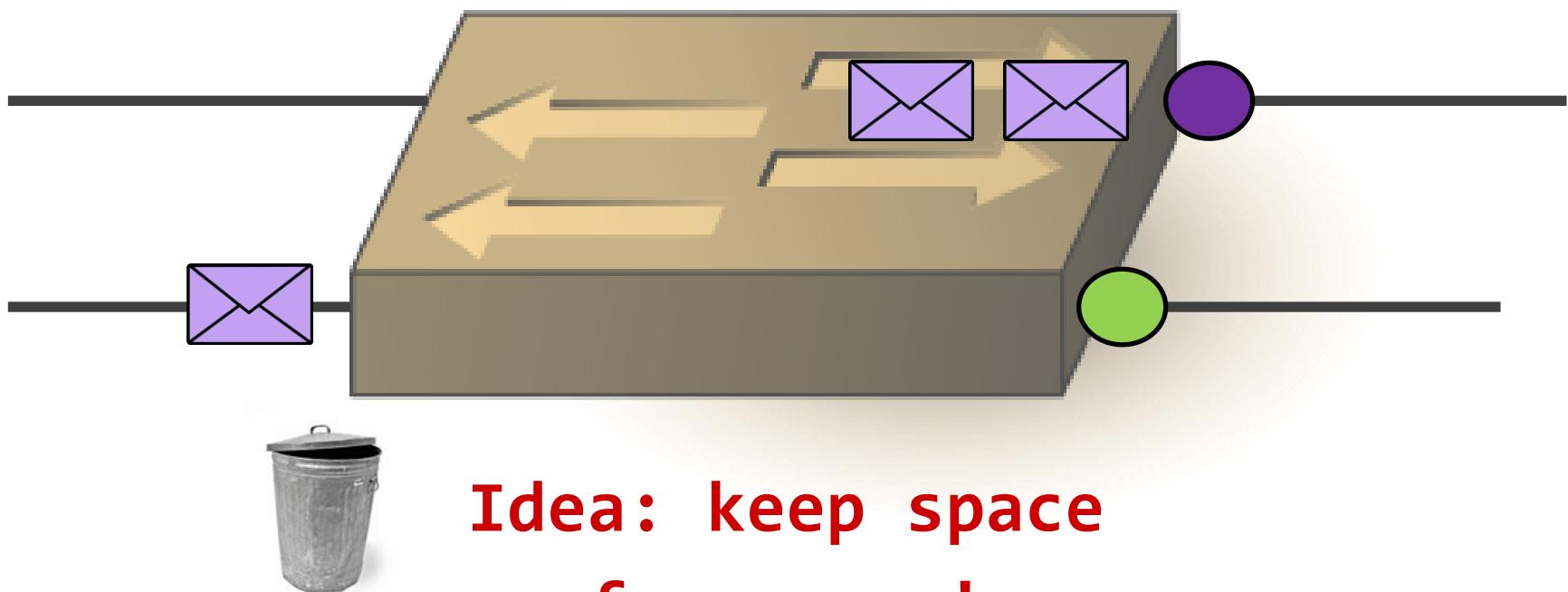
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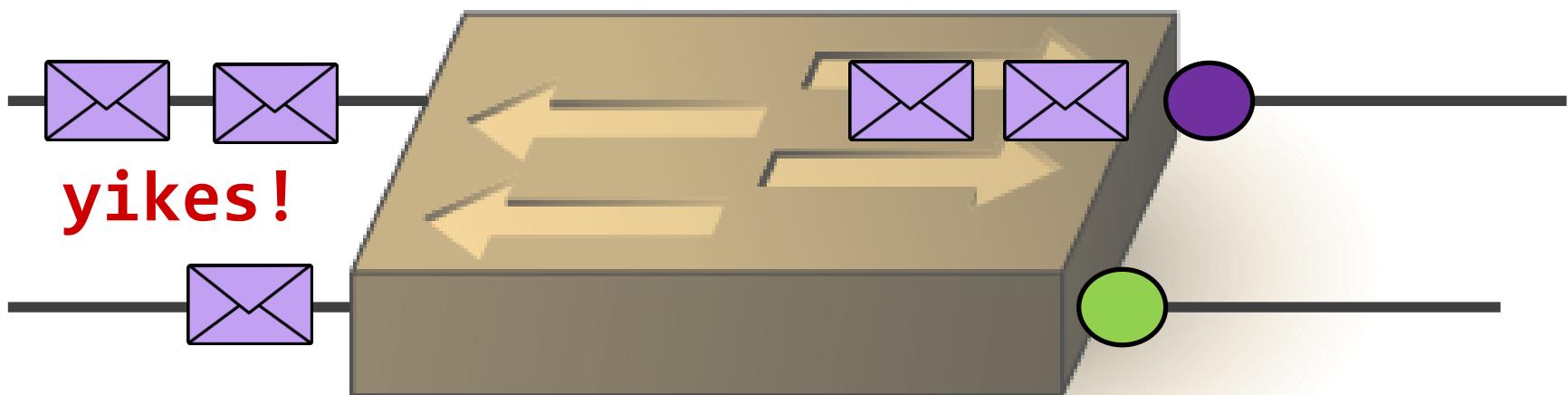


Idea: keep space
for green!



Scenario 2

→ What if switches become smart? Assume: shared memory size 3.



→ Suboptimal: drop to leave space but no space needed!

Credence

- Traffic at switch can be *predicted* fairly well
- AI/ML could significantly *improve buffer management*...
- ... and hence *admission control and throughput*!

Further reading:

[Credence: Augmenting Datacenter Switch Buffer Sharing with ML Predictions](#)

Vamsi Addanki, Maciej Pacut, and Stefan Schmid.

21st USENIX Symposium on Networked Systems Design and Implementation (**NSDI**), 2024.

Conclusion

- Opportunity: *structure* in demand and *reconfigurable* networks
- Enables self-driving networks
- Just the tip of the iceberg!
 - Optimal *collaboration* of ML, FM, and “human in the loop”?
 - Impact of self-driving layer on *other layers*?
 - *Scalable control* plane?
 - *Application-specific* self-adjusting networks,
e.g., for LLMs?



Online Video Course

Invitation to
Self-Adjusting Networks
A short video course

Before demand:

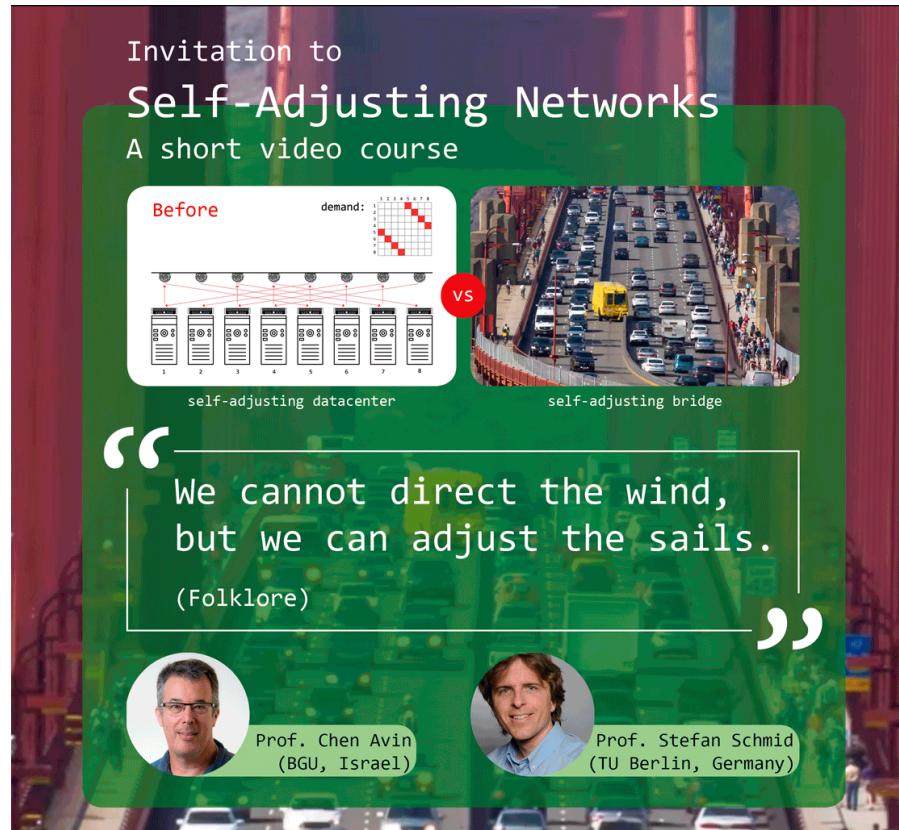
1	2	3	4	5	6	7	8
1	✓	✓	✓	✓	✓	✓	✓
2	✓	✓	✓	✓	✓	✓	✓
3	✓	✓	✓	✓	✓	✓	✓
4	✓	✓	✓	✓	✓	✓	✓
5	✓	✓	✓	✓	✓	✓	✓
6	✓	✓	✓	✓	✓	✓	✓
7	✓	✓	✓	✓	✓	✓	✓
8	✓	✓	✓	✓	✓	✓	✓

self-adjusting datacenter vs self-adjusting bridge

“ We cannot direct the wind,
but we can adjust the sails.
(Folklore) ”

Prof. Chen Avin
(BGU, Israel)

Prof. Stefan Schmid
(TU Berlin, Germany)



<https://self-adjusting.net/course>



YouTube Interview & CACM

Check out our **YouTube interviews**
on Reconfigurable Datacenter Networks:



[Revolutionizing Datacenter Networks via Reconfigurable Topologies](#)

Chen Avin and Stefan Schmid.

Communications of the ACM (CACM), 2025.

Watch here: <https://www.youtube.com/@self-adjusting-networks-course>



Websites

SELF-ADJUSTING NETWORKS
RESEARCH ON SELF-ADJUSTING DEMAND-AWARE NETWORKS

Project Overview Team Publications Contact Us

AdjustNet

Breaking new ground with demand-aware self-adjusting networks

Our Vision:
Flexible and Demand-Aware Topologies

Self-Adjusting Networks

new demand

+8+ routers

new flexible interconnect

MARCH 17, 2020

This site provides an overview of our ongoing research on the foundations of self-adjusting networks.

Download Slides

<http://self-adjusting.net/>
Project website



TRACE COLLECTION
WAN AND DC NETWORK TRACES

Publication Team Download Traces Contact Us

The following table lists the traces used in the publication: *On the Complexity of Traffic Traces and Implications*
To reference this website, please use: bibtex

File Name	Source Information	Type	Lines	Size	Download
exact_BoxLib_MultiGrid_C_Large_1024.csv	High Performance Computing Traces	Traces	17,947,800	151.3 MB	Download
exact_BoxLib_CNS_NoSpec_Large_1024.csv	High Performance Computing Traces	Traces	11,108,068	9.3 MB	Download
cesar_Nekbone_1024.csv	High Performance Computing Traces	Traces	21,745,229	184.0 MB	Download

<https://trace-collection.net/>
Trace collection website



Upcoming CACM Article

Revolutionizing Datacenter Networks via Reconfigurable Topologies

CHEN AVIN, is a Professor at Ben-Gurion University of the Negev, Beersheva, Israel

STEFAN SCHMID, is a Professor at TU Berlin, Berlin, Germany

With the popularity of cloud computing and data-intensive applications such as machine learning, datacenter networks have become a critical infrastructure for our digital society. Given the explosive growth of datacenter traffic and the slowdown of Moore's law, significant efforts have been made to improve datacenter network performance over the last decade. A particularly innovative solution is reconfigurable datacenter networks (RDCNs): datacenter networks whose topologies dynamically change over time, in either a demand-oblivious or a demand-aware manner. Such dynamic topologies are enabled by recent optical switching technologies and stand in stark contrast to state-of-the-art datacenter network topologies, which are fixed and oblivious to the actual traffic demand. In particular, reconfigurable demand-aware and "self-adjusting" datacenter networks are motivated empirically by the significant spatial and temporal structures observed in datacenter communication traffic. This paper presents an overview of reconfigurable datacenter networks. In particular, we discuss the motivation for such reconfigurable architectures, review the technological enablers, and present a taxonomy that classifies the design space into two dimensions: static vs. dynamic and demand-oblivious vs. demand-aware. We further present a formal model and discuss related research challenges. Our article comes with complementary video interviews in which three leading experts, Manya Ghobadi, Amin Vahdat, and George Papen, share with us their perspectives on reconfigurable datacenter networks.

KEY INSIGHTS

- Datacenter networks have become a critical infrastructure for our digital society, serving explosively growing communication traffic.
- Reconfigurable datacenter networks (RDCNs) which can adapt their topology dynamically, based on innovative optical switching technologies, bear the potential to improve datacenter network performance, and to simplify datacenter planning and operations.
- Demand-aware dynamic topologies are particularly interesting because of the significant spatial and temporal structures observed in real-world traffic, e.g., related to distributed machine learning.
- The study of RDCNs and self-adjusting networks raises many novel technological and research challenges related to their design, control, and performance.

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On the Complexity of Traffic Traces and Implications

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Chen Avin and Stefan Schmid.
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Cerberus: The Power of Choices in Datacenter Topology Design (A Throughput Perspective)

Chen Griner, Johannes Zerwas, Andreas Blenk, Manya Ghobadi, Stefan Schmid, and Chen Avin.
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Peter Gjøl Jensen, Morten Konggaard, Dan Kristiansen, Stefan Schmid, Bernhard Clemens Schrenk, and Jiri Srba.
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SyPer: Synthesis of Perfectly Resilient Local Fast Rerouting Rules for Highly Dependable Networks

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Matthew Nance Hall, Klaus-Tycho Foerster, Stefan Schmid, and Ramakrishnan Durairajan.

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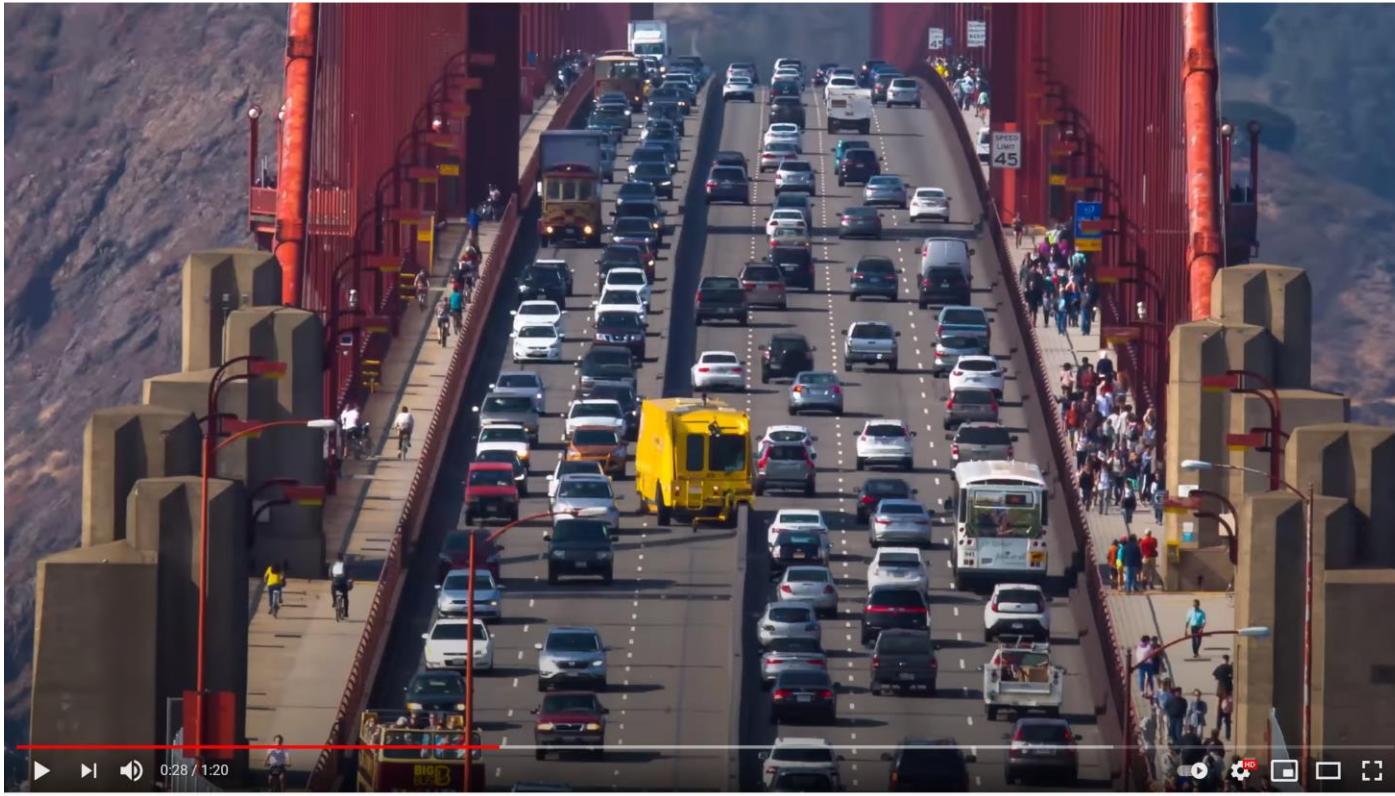
SplayNet: Towards Locally Self-Adjusting Networks

Stefan Schmid, Chen Avin, Christian Scheideler, Michael Borokhovich, Bernhard Haeupler, and Zvi Lotker.

IEEE/ACM Transactions on Networking (TON), Volume 24, Issue 3, 2016.

.

Questions?



Slides
available
here:

